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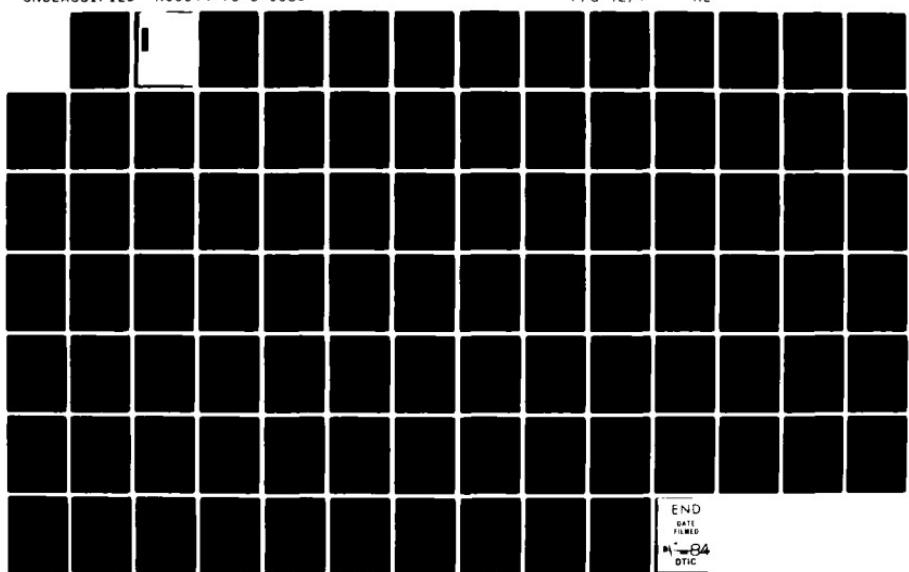
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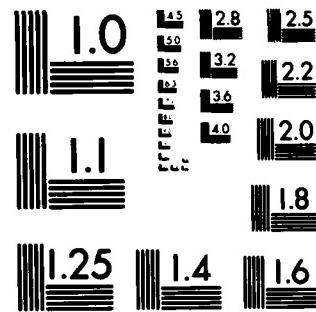


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THE SHAPLEY VALUE IN THE NON DIFFERENTIABLE CASE

by

Jean François Mertens

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THE SHAPLEY VALUE IN THE NON DIFFERENTIABLE CASE\*

by

J. F. Mertens \*\*

Introduction

In their book "Values of Non Atomic Games", Aumann and Shapley [1974] define the Shapley value for non atomic games, and prove existence and uniqueness of it for a number of important spaces of games like pNA and bv'NA. They also show that this value obeys the so-called diagonal formula, expressing the value of each infinitesimal player as his marginal contribution to the coalition of all players preceding him in a random ordering of the players. say if the worth  $v(S)$  of coalition  $S$  is expressed as a function of finitely many non atomic probabilities  $u_1, \dots, u_n$  by

$$v(S) = f(u_1(S), \dots, u_n(S)) , \quad f \in C^1 , \quad f(0) = 0$$

then the diagonal formula takes the form

$$[\phi(v)](S) = \sum_{i=1}^n u_i(S) \int_0^1 \frac{\partial f}{\partial x_i}(t, t, \dots, t) dt$$

or in general, more symbolically

$$[\phi(v)](ds) = \int_0^1 [v(tI + ds) - v(tI)] dt .$$

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This interpretation in terms of a random order depends on the fact that, for a large number of players, player  $\mu_s$  will in a random order occur at some time  $t$  uniformly distributed on  $[0,1]$ , and that the set of players preceding him will be an almost perfect sample of size  $t$  of the whole population - so that its worth will be essentially  $v(tI) = f(\mu_1(I), \dots, \mu_n(I))$ .

Those results have a large number of important applications - they do however depend on the differentiability of  $f$  along the diagonal.

The diagonal formula was later extended, in "Values and Derivatives", Mertens [1980], to a much wider class of games, including spaces like  $bv'NA$  in which the function  $f$  cannot be called differentiable.

The extended formula would apply say to majority games ( $v(S) = I(\mu(S) > \alpha) 0 < \alpha < 1$ ); or even to majority games in several different houses ( $v(S) = I(\mu_1(S) > \alpha_1, \mu_2(S) > \alpha_2, \dots, \mu_n(S) > \alpha_n) 0 < \alpha_i < 1$ ) provided all quota's  $\alpha_i$  are different. ( $I(\cdot)$  denotes the indicator function.)

But the case where the quotas would be the same - say all  $1/2$  - would be excluded, at least when  $n > 2$ .

Similarly, in economic applications, economies with strong complementarities, like "n-handed glove markets" ( $v(S) =$

$\min_{i=1, \dots, n} \mu_i(S)$ ) would remain excluded - again at least when  $n > 2$ .

Moreover, no value operator at all was known to exist on any space of games that would include all n-handed glove markets - except (Y. Tauman [1981]) when  $n$  is fixed and in addition all measures  $\mu_i$

are mutually singular, i.e., the different types of gloves have disjoint sets of owners.

S. Hart's "measure-based values" [1980] are an illuminating approach to this problem. They highlight the fact - which could already be seen in Aumann and Shapley's analysis [1974] of the three-handed glove market - that in some sense different finite approximations to the game may yield quite different values, according to one or another part of the player set - say the owners of one or another type of glove - approximates better the limiting game. For the approximations considered, the distribution of a random sample around the diagonal is essentially normal, with a covariance matrix that is quite sensitive to the relative degree of approximation in different parts of the player set.

Surprisingly, as we will show, in the limit the symmetry axiom - i.e., to ask that the solution depends only on the data of the game - is strong enough to force the distribution away from the normal distribution, and to impose, in some sense, a unique answer.

Here we extend the diagonal formula of Mertens [1980] to include, in addition, all situations of this type.

We get in this way a value - of norm 1 - on a closed space that will include DIFF - and DIAG -, the closed algebra generated by bv'NA, and also all games generated by a finite number of algebraic and lattice operations from a finite number of measures, and all markets functions of finitely many measures. The space will also include the finite games and the "regular" games with countably many players.

Intuitively, the diagonal formula is extended by taking the derivative not on the diagonal, but at some small perturbation of it - say  $tI + \epsilon X$  instead of  $tI$  - and by averaging the result for some probability distribution over perturbations. We prove further a weak form of uniqueness, in the sense that there is only one such probability distribution over perturbations that would yield a value.

In parallel, another extension is made to previous approaches, mainly in order to make the value invariant under all automorphisms of the lattice of coalitions, instead of only all automorphisms of the player set. In particular, this allows us to deal with finitely additive measures just as well as countably additive ones.

The basic definitions are given in Section 1, Section 2 defines the probability distribution over perturbations and shows its uniqueness. An explicit formula for the value of games of the type discussed above (n-handed glove markets, majority in several different houses) is derived in Section 3.

#### SECTION 1

We follow basically the terminology of Aumann and Shapley [1974].  $(I, C)$  denotes the player set,  $C$  being a  $\sigma$ -field of subsets of the set  $I$ . A game is a real valued function  $v$  on  $C$ , with  $v(\emptyset) = 0$ . Its variation norm  $\|v\|_{BV}$  is the supremum of the variation of  $v$  over all increasing chains  $(C_1 \subseteq C_2 \subseteq \dots \subseteq C_n)$  in  $C$ .  $(BV, \| \cdot \|_{BV})$  denotes the Banach algebra of all games of bounded variation.

FA is the subspace of BV consisting of additive set functions.

We are going to define a value - more precisely, a projection  $\Phi$  of norm 1 of some closed subspace Q of BV ( $FA \subseteq Q$ ) onto FA, such that  $[\Phi(v)](I) = v(I)$  and such that  $\Phi$  is symmetric in the sense that for any automorphism  $\theta$  of the Boolean algebra C, if  $\theta^t$  is defined on BV by  $[\theta^t(v)](S) = v(\theta(S))$ , then  $\theta^t(Q) = Q$  and  $\Phi \circ \theta^t = \theta^t \circ \Phi$ .

In fact,  $\Phi$  will be constructed as the composition of different positive linear symmetric mappings of norm 1:  $\Phi = \Phi_4 \circ \Phi_3 \circ \Phi_2 \circ \Phi_1$ .

(1.1)  $\Phi_1$  maps any game into the corresponding constant sum game,  $[\Phi_1(v)](S) = (1/2)[v(S) - v(S^C) + v(I)]$ : obviously  $\Phi_1$  is a symmetric projection of norm 1 onto the space  $Q_1$  of all constant sum games w ( $w(S) + w(S^C) = w(I)$ ), such that  $[\Phi_1(v)](I) = v(I)$ .

(1.2)  $\Phi_2$  is the extension operator:

$B(I, C)$  denotes the space of bounded measurable functions on  $(I, C)$ ,  $B_1^+(I, C)$  is the space of "ideal sets", i.e.,  $\{f | f \in B(I, C), 0 < f < 1\}$ .

For functions  $\bar{v}$  on  $B_1^+(I, C)$  (with  $\bar{v}(0) = 0$ ), one defines as previously the variation norm  $\|\bar{v}\|_{IBV}$  by considering all possible increasing chains in  $B_1^+(I, C)$ , and one defines  $\bar{v}^+(x) = \sup_{0 < f_j < f_{j+1} < x} \sum_i (\bar{v}(f_{i+1}) - \bar{v}(f_i))^+$ , and similarly for  $\bar{v}^-$ : obviously  $\|\bar{v}\|_{IBV} = \bar{v}^+(I) + \bar{v}^-(I)$ ,  $\bar{v} = \bar{v}^+ - \bar{v}^-$ .

Similar definitions are possible for the space  $F_\epsilon$  of functions  $\bar{v}$  ( $\bar{v}(0) = 0$ ) defined only on the  $\epsilon$ -neighborhood of the diagonal  $V_\epsilon = \{f \in B_1^+(I, C) : \sup(f) - \inf(f) < \epsilon\}$ , and lead to  $\|\bar{v}\|_{IBV, \epsilon}$  and

$\bar{v}_\epsilon^+$ ,  $\bar{v}_\epsilon^-$  by restricting all chains to remain in this neighborhood. By

definition  $\|\bar{v}\|_{IBV,0} = \inf_{\epsilon>0} \|\bar{v}\|_{IBV,\epsilon}$ .

Following Mertens [1981], we define  $F$  as the set of triplets  $(\pi, v, \epsilon)$ , where  $\pi$  is a finite measurable partition of  $I$ ,  $v$  is a finite set of non atomic elements of  $FA$  and  $\epsilon > 0$ .  $F$  is ordered by  $\alpha \leq \alpha'$  iff  $\pi_\alpha \leq \pi_{\alpha'}$  ( $\pi_{\alpha'}$  is a refinement of  $\pi_\alpha$ ) and  $v_\alpha \subseteq v_{\alpha'}$  and  $\epsilon_\alpha > \epsilon_{\alpha'}$ .  $(F, \leq)$  is filtering increasing.

$C_n$  is the set of increasing sequences  $0 \leq f_1 \leq f_2 \leq \dots \leq f_n \leq 1$  of measurable functions, and  $E_n$  the set of increasing sequences  $S_0 \subseteq S_1 \subseteq S_2 \dots \subseteq S_n$  in  $C$ .

For any  $\alpha \in F$  and  $f \in C_n$ , we define  $P_{\alpha,f}$  as the set of all probabilities with finite support on  $E_n$  such that  $\sum_i |E(I(S_i)) - f_i| < \epsilon$  uniformly on  $I$ , and such that  $S \in \pi_\alpha$ ,  $T \in \pi_\alpha$ ,  $S \cap T = \emptyset$  imply

(i)  $(S \cap S_i)_{i=1}^n$  is independent of  $(T \cap S_i)_{i=1}^n$

(ii)  $v_\alpha(S \cap S_i) = v_\alpha(f_i \cdot I(S))$

and (iii)  $f_i \cdot I(S) = 0 \Rightarrow S \cap S_i = \emptyset$ .

Intuitively  $P \in P_{\alpha,f}$  if  $P$  is the distribution of a random set (or sequence of sets) that is very similar to the ideal set  $f$  - "very" being measured by  $\alpha \in F$ .

Obviously  $\alpha \leq \alpha'$  implies  $P_{\alpha,f} \supseteq P_{\alpha',f}$ , and it follows from Mertens [1981] that always  $P_{\alpha,f} \neq \emptyset$ .

For any game  $v$ , and any  $f \in B_1^+(I, \mathcal{C})$ , let  $\bar{v}(f) = \lim_{\alpha \in F} \sup_{P \in P_{\alpha,f}} \int v(S) dP(S)$ ,  $\underline{v}(f) = \lim_{\alpha \in F} \inf_{P \in P_{\alpha,f}} \int v(S) dP(S)$ . (The inclusion

relations  $\alpha < \alpha' \Rightarrow P_{\alpha,f} \supseteq P_{\alpha',f}$  imply that the limits exist, the corresponding sup's or inf's being monotonic in  $\alpha$ .) Now  $v$  is in the domain  $D_\epsilon$  of  $\Phi_2^\epsilon$  iff  $\forall x \in V_\epsilon$ ,  $\bar{v}(x) = \underline{v}(x)$ , and then  $\Phi_2^\epsilon(v) \in F_\epsilon$  is defined by  $[\Phi_2^\epsilon(v)](x) = \bar{v}(x) = \underline{v}(x)$ .

Obviously  $D_\epsilon$  is a closed (in the maximum pseudo-metric) vector subspace of the space of all games, and symmetric, and  $\Phi_2^\epsilon$  is a symmetric linear operator from  $D_\epsilon$  to  $F_\epsilon$  [for the symmetry properties, it is sufficient to check that the set of non atomic elements of FA is invariant under any automorphism of the Boolean algebra C, and to define  $\theta(x)$  for  $x$  in  $B(I,C)$  in the obvious way if  $x$  is a step function, and by a uniform limit in general, so as to be able to define  $\theta^t$  on E].

Further  $\Phi_2^\epsilon$  transforms nonnegative games into nonnegative elements of  $F_\epsilon$ , and monotonic games into monotonic elements of  $F_\epsilon$ , and is of norm 1 both in the maximum norm and in the variation norm - this follows from the fact that  $\forall n, \forall f \in C_n, P_{\alpha,f} \neq \emptyset$ .

Finally one has obviously  $[\Phi_2^\epsilon(v)](1) = v(I)$ , and  $\epsilon_1 < \epsilon_2 \Rightarrow D_{\epsilon_1} \supset D_{\epsilon_2}$  and if  $v \in D_{\epsilon_2}$  then  $\Phi_2^{\epsilon_1}(v) = \Phi_2^{\epsilon_2}(v)$  on  $V_{\epsilon_1}$ .

Observe that for games with finitely many players,  $\Phi_2$  coincides with Owen's multilinear extension, and that for games in EXT (cfr. "Values and Derivatives"),  $\Phi_2$  coincides with the extension as defined in "Values and Derivatives".

Observe also that  $\Phi_2^\epsilon$  obviously maps constant sum games  $v \in D_\epsilon$  into constant sum games  $w \in F_\epsilon$  (i.e.:  $w(x) + w(1-x) = w(1) \forall x \in B_1^+(I,C)$ ), and that if  $v \in D_\epsilon$ , then  $\Phi_1(v) \in D_\epsilon$  and

$\phi_2^\epsilon \phi_1^\epsilon(v) = \phi_1^\epsilon \phi_2^\epsilon(v)$ , where  $\phi_1$  maps  $F_\epsilon$  into  $F_\epsilon$  by the formula  
 $[\phi_1(w)](x) = (1/2)[w(x) - w(1-x) + w(1)]$ .

(1.3)  $\phi_3$  is essentially the derivative operator defined in "Values and Derivatives":

First, if  $w \in F_\epsilon$ , define  $\bar{w}$  on  $\{f \in B(I, \mathbb{C}) : \sup f - \inf f < \epsilon\}$  by  $\bar{w}(f) = w[\max(0, \min(1, f))]$ .

Obviously  $w + \bar{w}$  is symmetric, positive, linear, etc., and if  $w$  is constant sum,  $\bar{w}(x) + \bar{w}(1-x) = \bar{w}(1)$ .

Let now  $[\phi_3(w)](x) = \lim_{\tau \rightarrow 0} \int_0^1 [\bar{w}(t + \tau x) - \bar{w}(t - \tau x)]/2\tau dt$   
whenever  $w \in \bigcup_\epsilon F_\epsilon$  and the limit exists for all  $x \in B(I, \mathbb{C})$ .

Some remarks are in order.

First, if one deals only with games in  $BV$ , there would be no problem of existence of the integrals - otherwise, we make the explicit assumption that, for any  $x$ ,  $\bar{w}(t + \tau x)$  is a.e. defined and integrable for all sufficiently small  $\tau$ .

We will also assume that, for all  $x \in B(I, \mathbb{C})$ ,  $\lim_{\tau \rightarrow 0} \int_0^1 [w(\tau(u+x)^+) + w(\tau(u+x)^-)] du = 0$ .

This is for instance satisfied, by the dominated convergence theorem, as soon as  $w$  is bounded and  $\lim_{\tau \rightarrow 0} w(\tau x) = 0 \forall x \in B_1^+(I, \mathbb{C})$ .

Obviously the mapping  $\phi_3$  is positive, linear, symmetric.

Let us show that

$$[\phi_3(w)](a + bx) = aw(1) + b[(\phi_3(w))(x)] \quad \forall a, b \in \mathbb{R}.$$

In particular we will have  $[\phi_3(w)](1) = w(1)$ , so that  $\phi_3(w)$  is linear on every plane containing the constants.

It is obvious that  $[\Phi_3(w)](bx) = b[\Phi_3(w)](x) \forall b, \forall x$ . So we only have to show that  $[\Phi_3(w)](1+x) = w(1) + [\Phi_3(w)](x)$ . Thus

$$\begin{aligned}\Phi_3(w)(1+x) &= \lim_{\tau \rightarrow 0} \int_0^1 \frac{\bar{w}(t+\tau+x) - \bar{w}(t-\tau-x)}{2\tau} dt \\ &= \lim_{\tau \rightarrow 0} \int_0^1 \frac{\bar{w}(t+\tau) - \bar{w}(t-\tau)}{2\tau} dt + \lim_{\tau \rightarrow 0} \int_0^1 \frac{\bar{w}(t+\tau+x) - \bar{w}(t-\tau-x)}{2\tau} dt \\ &\quad + \lim_{\tau \rightarrow 0} \frac{1}{2\tau} \int_0^1 [\bar{w}(t+\tau+x) - \bar{w}(t+\tau+x) - \bar{w}(t+\tau) - \bar{w}(t-\tau-x) \\ &\quad + \bar{w}(t-\tau+x) + \bar{w}(t-\tau)] dt.\end{aligned}$$

The first integral equals

$$\frac{1}{2\tau} \left[ \int_{-\tau}^{1+\tau} \bar{w}(s) ds - \int_{-\tau}^{1-\tau} \bar{w}(s) ds \right] = \frac{1}{2\tau} \left[ \int_{1-\tau}^{1+\tau} \bar{w}(s) ds - \int_{-\tau}^{\tau} \bar{w}(s) ds \right].$$

Since  $\bar{w}(x) + \bar{w}(1-x) = \bar{w}(1)$ , this equals

$$\bar{w}(1) - \frac{1}{2\tau} \left[ 2 \int_{-\tau}^{\tau} \bar{w}(s) ds \right] = w(1) - \frac{1}{\tau} \int_0^{\tau} w(s) ds = w(1) - \int_0^{\tau} w(\tau u) du,$$

and this last integral converges to zero by assumption. So the first integral converges to  $w(1)$ .

The second integral converges by definition to  $[\Phi_3(w)](x)$ , so there remains to show that the last integral converges to zero. This is equal to, writing  $F_+(t)$  for  $\bar{w}(t+\tau x) - \bar{w}(t)$ ,  $F_-(t)$  for  $\bar{w}(t-\tau x) - \bar{w}(t)$ :

$$\begin{aligned} & \frac{1}{2\tau} \left[ \int_0^1 (F_+(\tau + \tau) - F_+(\tau)) d\tau + \int_0^1 (F_-(\tau) - F_-(\tau - \tau)) d\tau \right] \\ &= \frac{1}{2\tau} \left[ \int_1^{1+\tau} F_+(s) ds - \int_0^\tau F_+(s) ds + \int_{1-\tau}^0 F_-(s) ds - \int_{-\tau}^0 F_-(s) ds \right]. \end{aligned}$$

Now the relation  $\bar{w}(x) + \bar{w}(1-x) = \bar{w}(1)$  implies

$F_+(\tau) = -F_-(1-\tau)$ , so that the last integral equals

$$\begin{aligned} &= \frac{1}{2\tau} \left[ - \int_{-\tau}^0 F_-(s) ds - \int_0^\tau F_+(s) ds - \int_0^\tau F_+(s) ds - \int_{-\tau}^0 F_-(s) ds \right] \\ &= \frac{-1}{\tau} \left[ \int_{-\tau}^0 F_-(s) ds + \int_0^\tau F_+(s) ds \right] \\ &= \frac{-1}{\tau} \int_0^\tau [F_+(s) + F_(-s)] ds \\ &= - \int_0^1 [\bar{w}(\tau(u+x)) - \bar{w}(\tau u) + \bar{w}(-\tau(u+x)) - \bar{w}(-\tau u)] du \\ &= - \int_0^1 [w(\tau(u+x)^+) + w(\tau(u+x)^-) - w(\tau u)] du \quad (\text{for } \tau < [1+|x|]^{-1}) \end{aligned}$$

and this last integral tends to zero by assumption.

Let us finally show that  $\Phi_3$  is of norm 1. Let  $x < x'$ , and consider any increasing chain

$$x < x_1 < x_2 < \dots < x_n < x' .$$

Denote by  $V(v)[x, x']$  the supremum of the variation of  $v$  over all such finite chains. Let  $\|x' - x\| = \delta$ , then  $V(v)[x, x'] < V(v)[x, x + \delta]$ , and there is no loss in restricting the chains to satisfy  $x_1 = x$ ,  $x_n = x + \delta$ .

If  $v = \Phi_3(w)$ , and we take  $\tau > 0$  sufficiently small such that all  $\bar{w}(t + \tau x_i)$  exist, then

$$\begin{aligned} \sum_i |v(x_{i+1}) - v(x_i)| &= \lim \frac{1}{2\tau} \sum_i \left| \int_0^1 [\bar{w}(t + \tau x_{i+1}) - \bar{w}(t + \tau x_i) \right. \\ &\quad \left. + \bar{w}(t - \tau x_i) - \bar{w}(t - \tau x_{i+1})] dt \right| \\ &< \lim \frac{1}{2\tau} \int_0^1 \left[ \sum_i |\bar{w}(t + \tau x_{i+1}) - \bar{w}(t + \tau x_i)| + \sum_i |\bar{w}(t - \tau x_i) \right. \\ &\quad \left. - \bar{w}(t - \tau x_{i+1})| \right] dt , \end{aligned}$$

or, letting  $|\bar{w}| = \overline{w_\epsilon^+} + \overline{w_\epsilon^-}$ ,

$$\begin{aligned} &< \lim \frac{1}{2\tau} \int_0^1 [|\bar{w}|(t + \tau(x + \delta)) - |\bar{w}|(t + \tau x) + |\bar{w}|(t - \tau x) - |\bar{w}|(t - \tau(x + \delta))] dt \\ &= \lim \frac{1}{2\tau} \left[ \int_1^{1+\tau\delta} |\bar{w}|(t + \tau x) dt + \int_{1-\tau\delta}^1 |\bar{w}|(t - \tau x) dt - \int_0^{\tau\delta} |\bar{w}|(t + \tau x) dt \right. \\ &\quad \left. - \int_{-\tau\delta}^0 |\bar{w}|(t + \tau x) dt \right] . \end{aligned}$$

But, for all  $x$ , we have  $0 < |\bar{w}|(x) < |\bar{w}|(1) = w_\epsilon^+(1) + w_\epsilon^-(1) = \|w\|_{IBV, \epsilon}$ . Thus

$$\sum_i |v(x_{i+1}) - v(x_i)| \leq \lim \frac{1}{2\tau} \left[ \int_{1-\tau\delta}^{1+\tau\delta} \|w\|_{IBV,\epsilon} dt - \int_{-\tau\delta}^{\tau\delta} (0) dt \right] \\ = \delta \|w\|_{IBV,\epsilon} ,$$

and therefore,  $\epsilon$  being arbitrary,

$$V(\Phi_3(w))[\chi, \chi'] \leq \|\chi' - \chi\| \cdot \|w\|_{IBV,0} .$$

In particular  $\|\Phi_3(w)\|_{IBV} = V(\Phi_3(w))[0,1] \leq \|w\|_{IBV,0} < \|w\|_{IBV}$ , which shows that  $\Phi_3$  is of norm 1.

Since  $v = \Phi_3(w)$  satisfies  $v(a + b\chi) = av(1) + bv(\chi)$ , we have

$$\frac{1}{2\tau} (v(t + \tau\chi) - v(t - \tau\chi)) = v(\chi) ,$$

so that  $(1/2\tau)[\bar{v}(t + \tau\chi) - \bar{v}(t - \tau\chi)] = v(\chi)$  for  $|\tau\chi| \leq t \leq 1 - |\tau\chi|$ .

If now  $\|v\|_{IBV} < \infty$ , then by  $v(a + \chi) = av(1) + v(\chi)$  we get

$V(v)[a - \delta, a + \delta] = V(v)[0, 2\delta]$ , which equals by homogeneity

$2\delta V(v)[0,1] = 2\delta \|v\|$ . Therefore, for all  $t$  we have

$$\left| \frac{1}{2\tau} [\bar{v}(t + \tau\chi) - \bar{v}(t - \tau\chi)] \right| \leq \|\chi\| \|v\| .$$

Also,  $\|v\| < \infty$  implies that  $\bar{v}(t + \tau\chi)$  is integrable (in  $t$ ) - as a function of bounded variation - so that, by Lebesgue's bounded convergence theorem

$$\int_0^1 \frac{1}{2\tau} [\bar{v}(t + \tau\chi) - \bar{v}(t - \tau\chi)] dt \rightarrow v(\chi) .$$

Thus, to show that  $v \in \text{Dom}(\Phi_3)$  and that  $\Phi_3(v) = v$  there only remains to show that  $v$  is bounded and  $\lim_{t \rightarrow 0} v(tx) = 0 \forall x \in B_1^+(I, C)$ . This follows again immediately from the boundedness of the norm of  $v$  and from the homogeneity.

Thus if  $v = \Phi_3(w)$ ,  $\|v\| < \infty$  implies  $v \in \text{Dom}(\Phi_3)$  and  $\Phi_3(v) = v$ . Therefore we get then  $v(v)[x, x'] = v(\Phi_3(w))[x, x'] \leq \|x' - x\| \cdot \|v\|$  and this relation is anyway true if  $\|v\| = +\infty$ . To summarize, we have shown that

Proposition 1:

- $\Phi_3$  is positive, linear, symmetric.
- $[\Phi_3(w)](1) = w(1)$ .
- $\|\Phi_3(w)\|_{IBV} \leq \|w\|_{IBV, 0}$ .
- $v \in \text{Range } (\Phi_3)$  implies
  - $v$  is linear on every plane containing the constants,
  - $v(v)[x, x'] \leq \|x' - x\| \cdot \|v\| \quad \forall x, x' \in B(I, C)$ ,
  - further, if  $\|v\| < \infty$ , then  $v \in \text{Dom } \Phi_3$ , and  $\Phi_3(v) = v$ .

For every  $\epsilon > 0$ , one gets a different domain for  $\Phi_1 \circ \Phi_2 \circ \Phi_3$ .

However, the composition having norm 1, and the domains being increasing when  $\epsilon \rightarrow 0$ , the composition can be extended to the closure of the union of those domains. Let us call  $\Psi$  this operator.

(In Section 4, we will show how to define directly an operator with closed domain extending  $\Psi$  - the present approach seems however easier for getting the main idea through - being more closely related to the literature.)

(1.4) Let us now prove part of our claims.

Let  $Q = \{v | v \in \text{Dom } \psi, \psi(v) \in \text{FA}\}$ : obviously  $Q$  is a closed symmetric space, and  $\psi$  is a value on  $Q$ .

It is obvious that  $Q$  contains DIFF, DIAG, and all games satisfying  $v(S) = v(S^C) \forall S \in C$ .

Let us show that  $Q$  also contains  $\text{bv}'\text{FA}$  (and all "regular" games with countably many players).

$\text{bv}'\text{FA}$  is the closed space generated by all games of the form  $v(S) = f(\mu(S))$ , where  $\mu \in \text{FA}_+^1$  and  $f$  is monotonic, continuous at zero and at 1, with  $f(0) = 0, f(1) = 1$ . It is sufficient to show that  $v \in Q$  when  $v$  is a generator. After applying  $\Phi_1$  to  $v$ , one may assume further that  $f(x) + f(1 - x) = 1$ .

Let us apply  $\Phi_2$ . Let  $\mu = \sum_{i>0} \mu_i$  where  $\forall i \mu_i > 0, i \neq j \Rightarrow \mu_i \neq \mu_j$ ,  $\mu_i$  is two-valued for  $i > 1$  and  $\mu_0$  is non atomic (i.e.,  $\sum_{i>1} \mu_i(I)$  is maximal given the other conditions). Assume without loss of generality that  $\mu_i(I)$  is monotonic in  $i > 1$ , and let  $a_i = \mu_i(I)$ ,  $v_i = a_i^{-1} \cdot \mu_i$  whenever  $a_i \neq 0$ ,  $n_a = \sup\{i | a_i \neq 0\}$ .

Denote by  $\pi_n$  any partition of  $I$  such that

$$\forall i, j \in \{1, \dots, n \wedge n_a\} \quad (i \neq j) \exists A \in \pi_n : \mu_i(A) \neq 0, \mu_j(A) = 0 .$$

Let  $\alpha_n = (\pi_n, \mu_0, 2^{-n})$ . Then for any  $P \in P_{\alpha_n, g}$  one has  $P$ -a.s.  $\mu_0(S) = \mu_0(g)$ , for  $i > 1$   $\mu_i(S) \in \{0, \mu_i(I)\}$  has expectation  $\mu_i(g)$  up to  $2^{-n}$  and  $\mu_1(S) \dots \mu_n(S)$  are independent.

Let also  $Y_g = \mu_0(g) + \sum_{i>1} a_i X_i$ , where the  $X_i$ 's are independent random variables with value in  $\{0, 1\}$ , and with expectation  $v_i(g)$ .

It follows that, when  $n$  goes to infinity, the distribution of  $\mu(S) (= \sum_{i>0} \mu_i(S))$  converges weakly to the distribution of  $Y_g$ . Further, when  $n_a < \infty$ , then  $\mu(S)$  is concentrated on the (finite) set of atoms of  $Y_g$ . It follows that  $E(v(S))$  converges to  $\bar{v}(g) = Ef(Y_g)$  except maybe when  $n_a = +\infty$  and the distribution of  $Y_g$  has some atoms at discontinuities of  $f$ .

Recall that, for  $g$  arbitrary, one sets  $\bar{v}(g) = \bar{v}(\max(0, \min(1, g)))$ . Obviously  $\bar{v}(t+g)$  is integrable - being monotonic. We now show that, even when  $n_a = +\infty$ ,  $\bar{v}(t+g)$  is the extension of  $v$  at  $t+g$ , except for at most two values of  $t$ . Indeed, the distribution of  $Y_g$  is obviously non atomic except when  $\lim_{i \rightarrow \infty} v_i(g) \wedge (1 - v_i(g)) = 0$ . Since we work only on some  $\epsilon$ -neighborhood of the diagonal, we can assume  $\sup(g) - \inf(g) < 1$ , so that the only possible exceptions occur when  $\lim_{i \rightarrow \infty} v_i[(t+g)^+] = 0$  and when  $\lim_{i \rightarrow \infty} v_i[(t+g) \wedge 1] = 1$ . It is sufficient to consider the first case, which is true for all  $t$  satisfying  $0 < t < -\limsup_{i \rightarrow \infty} v_i(g) = t_0$ . But if  $0 < t < -\limsup_{i \rightarrow \infty} v_i(g) = t_0$ , then  $B = \{\omega | g(\omega) < -(1/2)(t+t_0)\}$  is some measurable set, and, since  $v_i(B) = 0 \Rightarrow v_i(g) > -(1/2)(t+t_0)$ , one has  $v_i(B) = 1$  except at most finitely many times - otherwise one would have  $t_0 = -\limsup_{i \rightarrow \infty} v_i(g) < (1/2)(t+t_0)$ , thus  $t > t_0$  contrary to our assumption. Remark that on  $B$  one has  $(g+t)^+ = 0$ .

Thus, as soon as our partition  $\pi_a$  refines  $B$ , we will have that, with probability one,  $B \cap S = \emptyset$  i.e.,  $S \subseteq B^c$ ; and that  $v_i(B^c) > 0$  at most finitely many times - say  $v_i(B^c) = 0 \forall i > n_0$ . Therefore  $\mu(S)$  will have the distribution of  $v_0[(g+t)^+] + \sum_{i=1}^{n_0} a_i X_i$  where

$|E(x_i) - v_i((g+t)^+)| < 2^{-n}$ . Since  $n_0$  depends only on  $g$  and  $t$ , this implies  $\mu(S)$  is a distribution on a fixed, finite set of atoms, that converges weakly to the distribution of  $Y_{(g+t)^+}$  and thus the probability of every atom converges: we still have that  $\bar{v}(g+t)$  is the extension of  $v$  at  $(g+t)^+$ .

Thus the only possible troublesome value of  $t$  is  $t = -\limsup v_i(g)$  (and dually  $1-t = \liminf v_i(g)$ ).

In particular, for any  $x$ ,  $\bar{v}(t+\tau x)$  is a.e. defined and integrable for all sufficiently small  $\tau$ . The second condition for  $v \in \text{Dom } \Phi_3$  was satisfied as soon as  $v$  is bounded and  $\bar{v}(\tau x)$  converges to zero for all  $x \in B_1^+(I, \mathbb{C})$ ;  $v$  being monotonic it is sufficient to show that  $\lim_{\tau \downarrow 0} \bar{v}(\tau) = 0$ ; this follows from  $\bar{v}(\tau) \leq (\text{Cav } f)(\tau)$  because  $\text{Cav } f$  is continuous and vanishes at zero,  $f$  having this property.

Thus to show that  $v \in Q$  there only remains to show that  $\frac{1}{2\tau} \int_0^1 [\bar{v}(t+\tau x) - \bar{v}(t-\tau x)] dt$  converges to some element of  $FA$ . To facilitate this, we begin by the lemma.

Lemma: If

- $\bar{v}(t+\tau x)$  is, for every  $x$ , a.e. defined and integrable for all sufficiently small  $\tau$
  - $\int_0^1 [v[\tau(t+x)^+] + v[\tau(t+x)^-]] dt$  converges to zero with  $\tau (> 0)$  for all  $x$
  - $\frac{1}{2\tau} \int_0^1 [\bar{v}(t+\tau x) - \bar{v}(t-\tau x)] dt$  converges for all  $x \in B_1^+(I, \mathbb{C})$
- then  $v \in \text{Dom } \Phi_3$  - i.e., the last expression converges for all  $x \in B(I, \mathbb{C})$ .

Proof: Our assumption immediately implies the convergence when  $x < 0$ , and it is sufficient to prove the convergence for any  $\psi$  satisfying  $\psi < 1$ . Write thus  $\psi = 1 + x$ , with  $x < 0$ ; the computation we did when proving that  $\Phi_3(w)$  is linear on every plane containing the constants proved also that  $\Phi_3(w)(\psi)$  exists - and this finishes the proof.

Q.E.D.

By virtue of this lemma, it is sufficient to show that  $\phi(\tau, x) = \frac{1}{(1/2)\tau} \int_0^1 [\bar{v}(t + \tau x) - \bar{v}(t - \tau x)] dt$  converges to some element of FA for all  $x \in B_1^+(I, C)$ , where  $\bar{v}(g) = Ef(Y_{\bar{g}})$ , with  $\bar{g} = 0 \vee (g \wedge 1)$ , and  $Y_g = a_0 v_0(g) + \sum_{i=1}^{n_a} a_i X_i$ , where the random variables  $X_i$  are independent, with values in  $\{0, 1\}$  and with expectation  $v_i(g)$ .

Let now  $f_L(x) = \sum_{y < x} (f(y+) - f(y))$  where  $f(y+) = \lim_{\epsilon \downarrow 0} f(y + \epsilon)$ . Let also  $f_R(x) = f(x) - f_L(x)$ : then both  $f_L$  and  $f_R$  are increasing,  $f_L$  is left continuous and  $f_R$  is right continuous, so that

$$f(x) = \int_0^1 I(x > q) df_R(q) + \int_0^1 I(x > q) df_L(q) .$$

Since  $\phi(\tau, x)$  depends linearly on  $f$ , and since  $0 < \phi(\tau, x) < \phi(\tau, 1) < 1$  using the monotonicity of  $v$  and the relation  $0 < x < 1$ , we can apply Fubini's theorem to get

$$\phi(\tau, x) = \int_0^1 \phi_{q,R}(\tau, x) df_R(q) + \int_0^1 \phi_{q,L}(\tau, x) df_L(q)$$

when  $\phi_{q,R}$  and  $\phi_{q,L}$  denote the function  $\phi$  corresponding to the case where  $f(x) = I(x > q)$  and  $f(x) = I(x > q)$  respectively.

But  $\phi_{q,R}(\tau, x)$  and  $\phi_{q,L}(\tau, x)$  being uniformly bounded, the bounded convergence theorem implies that it is sufficient to prove that  $\phi_{q,R}(\tau, x)$  and  $\phi_{q,L}(\tau, x)$  converge to some element of FA.

Thus, we have reduced the problem to the case where  $f(x) = I(q < \cdot x)$ , where  $<\cdot$  stands for either  $<$  or  $\leq$ .

Remark that for  $i \geq 1$ ,  $v_i(\bar{g}) = 0 \vee (v_i(g) \wedge 1)$ , and that  $v_i(t + \tau x) = t + \tau v_i(x)$ . Let for short  $p_i = v_i(x)$  (thus  $0 < p_i < 1$ ), and let  $Z_i$  ( $i = 1, \dots, n_a$ ) be independent random variables, uniformly distributed over  $[0, 1]$ . Then

$$\begin{aligned}\phi(\tau, x) &= \frac{1}{2\tau} \int_0^1 E[I[a_0 v_0((t - \tau x)^+)] + \sum_{i=1}^{n_a} a_i I(Z_i < t - \tau p_i)] < q \\ &\quad < a_0 v_0((t + \tau x) \wedge 1) + \sum_{i=1}^{n_a} a_i I(Z_i < t + \tau p_i)] dt\end{aligned}$$

where  $<\cdot$  stands for  $<$  or  $\leq$  when  $<\cdot$  is  $<$  or  $\leq$  respectively.

Let also  $\bar{\phi}(\tau, x)$  be the same expression, with  $v_0[(t - \tau x)^+]$  replaced by  $t - \tau p_0$  and  $v_0[(t + \tau x) \wedge 1]$  replaced by  $t + \tau p_0$ . Then obviously  $\bar{\phi} \geq \phi$ , and the integrands can differ only when  $t < \tau$  or  $t \geq 1 - \tau$ , so that

$$\begin{aligned}\bar{\phi}(\tau, x) - \phi(\tau, x) &\leq \frac{1}{2\tau} \int_0^\tau E[I[q < a_0(t + \tau p_0) + \sum_{i=1}^{n_a} a_i I(Z_i < t + \tau p_i)] dt \\ &\quad + a \text{ similar integral between } 1 - \tau \text{ and } \tau.\end{aligned}$$

Obviously, the right hand member goes to zero with  $\tau$ . Thus if we set

$$\begin{aligned}\psi(p) = \frac{1}{2} \int_0^1 E[I[a_0(t - p_0) + \sum_{i>1} a_i I(Z_i < t - p_i) <: q <: a_0(t + p_0)] \\ + \sum_{i>1} a_i I(Z_i > t + p_i)] dt\end{aligned}$$

we have to prove that  $\psi$  is differentiable at zero, i.e., that  $\lim_{\tau \rightarrow 0} (1/\tau)\psi(\tau p)$  exists and is linear in  $p$  - i.e., a continuous linear functional on  $\ell_\infty$ . We will even show that the limit is of the form  $\sum \gamma_i p_i$ , with  $\gamma_i > 0$ ,  $\sum \gamma_i = 1$ .

We will show that this is the case for all sequences  $p_i$  having only finitely many non zero terms.

The result will follow from this, because, for an arbitrary sequence  $p_i$  ( $0 < p_i < 1$ ), one has then by monotonicity

$$\liminf_{\tau \rightarrow 0} \frac{1}{\tau} \psi(\tau p) \geq \sum_{i=0}^k \gamma_i p_i , \text{ thus}$$

$$\liminf_{\tau \rightarrow 0} \frac{1}{\tau} \psi(\tau p) \geq \sum_0^\infty \gamma_i p_i , \text{ and}$$

$$\liminf_{\tau \rightarrow 0} \frac{1}{\tau} \psi(\tau(1-p)) \geq 1 - \sum_0^\infty \gamma_i p_i$$

and since  $(1/\tau)[\psi(\tau p) + \psi(\tau(1-p))]$  converges to 1 (this is the computation we did when proving that any  $w$  in the range of  $\Phi_3$  is linear on every plane containing the constants), it follows that

$$\lim_{\tau \rightarrow 0} (1/\tau)\psi(\tau p) = \sum_0^\infty \gamma_i p_i .$$

Thus we have to show that  $\psi(p)$  is differentiable at zero as a function of the variables  $p_0 \dots p_k$ , the other  $p_i$ 's being fixed at zero.

Further if then  $\gamma_i = (\partial\psi/\partial p_i)_{p=0}$ , we have to show that  $\sum_{i>0} \gamma_i = 1$  (because obviously  $\gamma_i > 0$  by monotonicity of  $\psi$ ). Writing the expectation in the formula of  $\psi(p)$  as the expectation of the conditional expectation given  $Z_1 \dots Z_k$  yields

$$\psi(p) = \frac{1}{2} \sum_{y \in \{0,1\}^k} \int_0^1 [H_t^y(p) - H_t^y(-p)] dt ,$$

where

$$H_t^y(p) = \prod_{i=1}^k \left( \left[ \frac{1}{2} + (2y_i - 1)(t - p_i - \frac{1}{2}) \right]^+ \wedge 1 \right) F_t^k(q + a_0 p_0 - \sum_{i=1}^k a_i y_i) ,$$

with

$$F_t^k(x) = P(a_0 t + \sum_{i>k} a_i I(Z_i < t) <: x) .$$

Let

$$\bar{H}_t^y(p) = \prod_{i=1}^k \left[ \frac{1}{2} + (2y_i - 1)(t - p_i - \frac{1}{2}) \right] F_t^k(q + a_0 p_0 - \sum_{i=1}^k a_i y_i) ,$$

and

$$\bar{\psi}(p) = \frac{1}{2} \sum_{y \in \{0,1\}^k} \int_0^1 [\bar{H}_t^y(p) - \bar{H}_t^y(-p)] dt .$$

One shows, just as before for  $\Phi - \phi$ , that  $\bar{\psi} - \psi$  is differentiable at zero with zero differential (the difference of the integrands is anyway small, and different from zero only on a small part of the domain).

Thus to show the differentiability at zero, it is sufficient to

show the differentiability at zero of the expression

$$\left( \prod_{i=1}^k p_i^{n_i} \right) \int_0^1 t^n F_t^k(x \pm a_0 p_0) dt$$

since  $\bar{\Psi}(p)$  is a linear combination of expressions of this type.

Since  $\left( \prod_{i=1}^k p_i^{n_i} \right)$  is obviously differentiable, this amounts in turn to the differentiability at zero of  $\int_0^1 t^n F_t^k(x \pm a_0 p_0) dt$ .

If this differentiability is proved, then using  $k = 1$ ,

$$\begin{aligned} r_1 &= \lim_{p_1 \rightarrow 0} \frac{1}{p_1} \bar{\Psi}(p_1) = \lim_{p_1 \rightarrow 0} \frac{1}{2p_1} \int_0^1 2p_1 [F_t^1(q) - F_t^1(q - a_1)] dt \\ &= \int_0^1 [F_t^1(q) - F_t^1(q - a_1)] dt \\ &= E \int_0^1 [a_0 t + \sum_{i>1} a_i I(z_i < t) <: q <: a_0 t + a_1 + \sum_{i>1} a_i I(z_i < t)] dt \end{aligned}$$

or, since  $Z_1$  is, like  $t$ , uniformly distributed on  $[0,1]$  and independent of the other  $Z_i$ 's, we get

$$r_1 = P[a_0 Z_1 + \sum_{i>1} a_i I(z_i < z_1) <: q <: a_0 Z_1 + \sum_{i>1} a_i I(z_i < z_1)] .$$

Let, for  $k > 1$ ,  $J_k(\omega)$  denote the random interval (of length  $a_k$ )

$$\{x | a_0 Z_k + \sum_{i>1} a_i I(z_i < z_k) <: x <: a_0 Z_k + \sum_{i>1} a_i I(z_i < z_k)\}$$

(obviously the  $J_k(\omega)$  are disjoint if we restrict ourselves to the set of  $\omega$ 's (with probability one) where  $i \neq j \Rightarrow z_i(\omega) \neq z_j(\omega)$ ,  $0 < z_i(\omega) < 1$ ).

Let also  $J_o(\omega) = [0,1] \cup \bigcup_{k>1} J_k(\omega)$ . In those notations

$$\gamma_i = \left( \frac{\partial \psi}{\partial p_i} \right)_{p=0} = P(q \in J_i(\omega)) \text{ for all } i > 1 .$$

Similarly, using  $k = 0$ , we get, using  $w(t) = a_0 t + \sum_{i>1} a_i I(Z_i < t)$ ,

$$\begin{aligned} \gamma_0 &= \lim_{p_0 \rightarrow 0} \frac{1}{p_0} \bar{\psi}(p_0) = \lim_{p_0 \rightarrow 0} \frac{1}{2p_0} \int_0^1 [F_t^0(q + a_0 p_0) \\ &\quad - F_t^0(q - a_0 p_0)] dt \\ &= \lim_{p_0 \rightarrow 0} \frac{1}{2p_0} \int_0^1 P[q - a_0 p_0 < w(t) < q + a_0 p_0] dt . \end{aligned}$$

Thus, if  $a_0 = 0$ , then  $\gamma_0 = 0$ , and the differentiability condition to check is obvious, so there only remains to show that  $\sum_{i>1} \gamma_i = P(q \in \bigcup_i J_i(\omega)) = 1$ . When there are only finitely many non zero  $a_i$ 's, then  $q \in \bigcup_i J_i(\omega) = \{x | 0 < x < 1\}$  for any  $\omega$ , while if there are countably many non zero  $a_i$ 's, a recent result of Berbee [1981] proves that  $P(q \in \bigcup_i J_i(\omega)) = 1$  [even that  $P(q \in \bigcup_i J_i(\omega)) = 1$ ].

There remains therefore to consider the case  $a_0 > 0$ . Since

$1 - \sum_{i>1} \gamma_i = P(q \in J_o(\omega))$ , the property  $\sum_i \gamma_i = 1$  amount to

$$P(q \in J_o(\omega)) = a_0 \lim_{p_0 \rightarrow 0} \frac{1}{2a_0 p_0} \int_0^1 P[q - a_0 p_0 < w(t) < q + a_0 p_0] dt .$$

On the other hand, the differentiability at zero of  $\int_0^1 t^n F_t^k(x + a_0 p_0) dt$ , when  $F_t^k(x) = P(a_0 t + \sum_{i>k} a_i I(Z_i < t) < x)$  can be rewritten, by letting  $a'_i = a_{k+i}$  for  $i > 1$ ,  $a'_0 = a_0$ ,  $\sigma = \sum_{i>0} a'_i$  ( $a_0 > 0 \Rightarrow \sigma > 0$ ),  $a''_i = a'_i/\sigma$ ,

$F_t(y) = F_t^k(ay)$  ( $= P(a''t + \sum_{i=1}^{n-1} a_i'' I(z_i < t) <: y)$ ), as the differentiability at zero of  $\int_0^t \int_{z_i}^y F_t(\frac{x}{\sigma} + a'' p_o) dt$  or, letting  $z = x/\sigma$ , and writing  $a_i$  for  $a_i''$  - so  $F_t$  becomes  $F_t^o$  - as the differentiability of  $a_o \int_0^t \int_{z_i}^y F_t^o(z) dt = \Phi_n(z)$  as a function of  $z$ .

To show also that  $\sum \gamma_i = 1$ , we have to show further that, when  $n = 0$ , the derivative is  $P(z \in J_o(\omega))$ . We have

$$\frac{1}{\delta} [\Phi_n(z + \delta) - \Phi_n(z)] = E \frac{a_o}{\delta} \int_0^t \int_{z_i}^y I[z < w(t) < z + \delta] dt .$$

Let  $T_z = \inf \{t > 0 : z < w(t)\}$ : if  $z < w(t) < z + \delta$ , we have  $T_z < t < T_z + \delta/a_o$ , thus  $T_z^n < t^n < T_z^n + [(1 + \delta/a_o)^n - 1]$ , therefore, if  $X = (a_o/\delta) \int_0^t \int_{z_i}^y I[z < w(t) < z + \delta] dt$ ,

$$\begin{aligned} X &< \frac{a_o}{\delta} \int_0^t \int_{z_i}^y I(z < w(t) < z + \delta) dt < X + \frac{a_o}{\delta} [(1 + \frac{\delta}{a_o})^n - 1][(T_z + \frac{\delta}{a_o}) - T_z] \\ &= X + (1 + \frac{\delta}{a_o})^n - 1 \end{aligned}$$

Now  $X = \int_{z_i}^y I_{J_o(\omega)}(z + \delta u) du$ .

If  $z \in \cup J_i(\omega)$ , then  $\lim_{x \rightarrow z} I_{J_i(\omega)}(x) = 1$ , except maybe if  $z$  is a boundary point of some  $J_i(\omega)$  - but this event has probability zero,

even conditionally on all  $Z_j$  ( $j \neq i$ ) (using  $a_o > 0$ ).

If  $z \in J_o(\omega)$ , then  $(1/\delta) \int_z^{z+\delta} I_{J_o(\omega)}(x) dx = 1 - \sum_{i=1}^{\infty} (1/\delta) \int_z^{z+\delta} I_{J_i(\omega)}(x) dx$ , and it is sufficient to show that the conditional expectation (given  $z \in J_o(\omega)$  and given  $T_z$ ) of the sum converges to zero. Now, if  $z \in J_o(\omega)$ ,  $x > z$ , then  $I_{J_i(\omega)}(x) <$

$I(T_{z \vee (x-a_i)} < z_i < T_x) \leq I(T_{z \vee (x-a_i)} < z_i < T_{z \vee (x-a_i)} + ((x - z) \wedge a_i)/a_0)$ , and then

$$P(x \in J_i(\omega) | T_z, z \in J_o(\omega)) \leq \frac{1}{1 - T_z} \left[ \frac{(x - z) \wedge a_i}{a_0} \right]$$

Thus

$$P(x \in \bigcup_i J_i(\omega) | z \in J_o(\omega), T_z) \leq \frac{1}{a_0(1 - T_z)} \sum_{i \geq 1} [(x - z) \wedge a_i].$$

Also, for  $a_0 > 0, 1 - T_z > 0$  with probability one if  $z < 1$ , and since  $\sum a_i < \infty$ , it follows that the right hand member goes to zero when  $x + z < 1$ . Thus the left hand member being bounded, we get, if  $z < 1$ , and obviously also if  $z \geq 1$ ,  $\lim_{x \rightarrow z} P(x \in \bigcup_i J_i(\omega), z \notin \bigcup_i J_i(\omega)) = 0$ ; and therefore by symmetry  $\lim_{x \rightarrow z} P(x \in \bigcup_i J_i(\omega), z \in \bigcup_i J_i(\omega)) = 0$  and thus  $I_{\bigcup_i J_i(\omega)}(x)$  is continuous in probability. In particular  $\sum_{i \geq 1} \gamma_i(x)$  is a continuous function of  $x$ , and also  $T_z^n I_{J_o(\omega)}(x)$  converges in probability to  $T_z^n I_{J_o(\omega)}(z) I(0 < x < 1)$  so that by the bounded convergence theorem  $(1/\delta) [\phi_n(z + \delta) - \phi_n(z)]$  converges to  $I(z + \delta \in [0, 1]) E[T_z^n I_{J_o(\omega)}(z)]$ .

Since the equation  $\Phi'(z) = P(z \in J_o(\omega))$  is needed only for  $0 < z < 1$ , we have proved our statement. (Remark that the differentiability condition of  $\int_0^1 t^n F_t^k(x \pm a_0 p_0) dt$  at zero was only one-sided since  $a_0 p_0 > 0$ .)

Remark 1: A closer look at the above argument shows that in fact we proved more: if  $\phi(x) = (1/2) \int_0^1 [\tilde{v}(t + x) - \tilde{v}(t - x)] dt$ , then

$\phi(x)$  is Frechet-differentiable at zero. Indeed, the proof of the lemma shows that it is sufficient to consider  $x \in B_1^+(I, \mathbb{C})$  - provided  $\int_0^1 (v[\tau(u + x)^+] + v[\tau(u + x)^-]) du$  converges to zero uniformly over the unit ball, which is obvious whenever  $v$  is norm continuous at zero.

Similarly the bounded convergence theorem still permits the reduction to the case where  $f(x) = I(q < x)$ . Also the approximation of  $\phi$  by  $\bar{\phi}$  and later of  $\psi$  by  $\bar{\psi}$  are obviously uniform in  $p \in [0,1]^\infty$ .

Since, as we just mentioned, the convergence of  $(1/\tau)[\psi(\tau p) + \psi(\tau(1-p))]$  to 1 is uniform in  $p$  for  $v$  norm continuous at zero, it will be sufficient to consider vectors  $p$  such that  $p_i = 0 \forall i > k$ : indeed, the same conclusion will then hold when  $p_i = 1 \forall i > k$ , so that if  $k$  is chosen such that  $\sum_{i>k} y_i < \epsilon$ , then  $\tau_0$  such that,  $\forall \tau: |\tau| < \tau_0, \forall p$  in one of these two classes,

$| (1/\tau)\psi(\tau p) - \sum y_i p_i | < \epsilon$ , the result will follow (from the monotonicity of  $\bar{\psi}$ ) for arbitrary  $p \in [0,1]^\infty$  by sandwiching it between the two approximations  $p_i, \bar{p}_i$ , where  $p_i = \bar{p}_i = p_i$  for  $i \leq k$  and for  $i > k$ ,  $p_i = 0, \bar{p}_i = 1$ . As shown in the proof, the differentiability of  $\bar{\psi}$  over  $p$ 's having only  $k$  non zero coordinates amounts to the differentiability of a product of functions of 1 variable, which is true as soon as each factor in the product is differentiable, what we proved.

Also we did not need in fact the symmetry of  $\phi$ . We thus obtain finally:

Proposition 2: Let

$$H(x) = \int_0^1 \bar{v}(t + x) dt.$$

Then  $H$  is Frechet differentiable at zero, with as derivative the value of  $v$ :

$$\int_0^1 \sum_{i=0}^{\infty} \gamma_i(x) v_i(\cdot) df(x)$$

where in the integration a discontinuity to the right (left) of  $x$  is to be interpreted as the corresponding mass immediately to the right (left) of  $x$ .

The  $\gamma_i(x)$  are defined in the following way. Assume the game  $v$  is of the form  $v(s) = f(\mu(s))$ , where  $\mu = \sum_{i=0}^{\infty} a_i v_i$ ,  $a_i > 0$ ,  $v_i > 0$ ,  $v_i(1) = 1$ ,  $\sum_i a_i = 1$ ,  $v_0$  non atomic and  $i > 1 \Rightarrow v_i$  two valued,  $i \neq j \Rightarrow v_i \neq v_j$ . Define random variables  $Z_i$  independent and uniformly distributed over  $[0,1]$ , then expand each point  $z_i$  to some open interval of length  $a_i$ , then shrink the remaining part of  $[0,1]$  (of length 1) to length  $a_0$  (proportionately). Denote by  $J_i$  the random interval thus obtained corresponding to  $Z_i$ . Then  $\gamma_i(x) = P(x \in J_i)$  for  $i > 1$ , and  $\gamma_0(x) = P(\bigcup_i J_i \text{ has density 0 at } x)^{1/}$

$$[\text{i.e. } \gamma_0(x) = P\left(\limsup_{\delta \rightarrow 0} \frac{1}{2\delta} \int_{x-\delta}^{x+\delta} I_{\bigcup_i J_i}(x) dx = 0\right)]$$

When there are infinitely many players, we also showed that the  $\gamma_i(x)$ , ( $i = 0, 1, \dots$ ) are continuous on  $[0,1]$ , with  $\gamma_0(0) = \gamma_0(1) = 1$  if  $a_0 > 0$  - in particular, if  $a_0 > 0$ , the series  $\sum_i \gamma_i$  is uniformly convergent (to 1) on  $[0,1]$ , and anyway  $\sum_i \gamma_i(x) = 1 \forall x: 0 < x < 1$ .

In particular, when there are infinitely many players, the value

of  $f(x)$  at a jump and the exact definition of the  $\int \dots df(x)$  play no role.

It is possible to draw still some sharper conclusions from the foregoing: let  $\mu^n \in FA_+^1$  converge (in norm) to  $\mu^0 \in FA_+^1$ . Let  $v_i$  ( $i > 1$ ) enumerate all atoms of all  $\mu^n$  and let  $v_o^n$  be the non atomic part of  $\mu^n$ , with  $\mu^n = a_o^n v_o^n + \sum_{i=1}^{\infty} a_i^n v_i$ . One sees immediately that  $a_o^n v_o^n$  is norm convergent to  $a_o^0 v_o^0$ , and that  $(a_i^n)_{i=0}^{\infty}$  is  $\ell_1$  - convergent to  $(a_i^0)_{i=0}^{\infty}$ . Assume now  $\mu^0$  has infinitely many players. Since this implies that when realizing the random ordering with the same set of random variables  $Z_i$ , we will have a.s.  $J_i^n \rightarrow J_i^0$ , and since  $\gamma_i(q) = P(q \in J_i) = P(q \in J_i^n)$ , it will follow that  $\forall i > 1$ ,  $\gamma_i^n(q_n) \rightarrow \gamma_i^0(q_o)$  whenever  $q_n \rightarrow q_o$ , ( $0 < q_o < 1$ ). But  $\gamma_i^n(q) < \text{Prob}(q - a_i^n < W_n(Z_i) < q) < a_i^n/a_o^n$ , so that  $\sum_{i>k} \gamma_i^n(q) < (1/a_o^n) \sum_{i>k} a_i^n$  if  $a_o^n > 0$ . Thus the  $\ell_1$  - convergence of  $a_i^n$  to  $a_i^0$  implies that, if  $a_o^0 > 0$ ,

$\limsup_{k \rightarrow \infty} \sup_{n \in \mathbb{N}} \sum_{i>k} \gamma_i^n(q) = 0$ , i.e., the convergence of the series  $\sum_{i>1} \gamma_i^n(q)$  is uniform in  $n$  and  $q$ . Since  $\gamma_i^n(q_n) \rightarrow \gamma_i^0(q_o)$ , it follows that  $\sum_{i>1} \gamma_i^n(q_n) \rightarrow \sum_{i>1} \gamma_i^0(q_o)$ , and the relation  $\sum_{i>0} \gamma_i = 1$  yields therefore  $\gamma_o^n(q) \rightarrow \gamma_o^0(q)$ . If  $a_o^0 = 0$ , then  $\gamma_o^0 = 0$  and therefore  $\liminf_n \gamma_o^n(q_n) > \gamma_o^0(q_o)$ , so that the relations  $\gamma_i^n > 0$ ,  $\sum_{i=0}^{\infty} \gamma_i^n = 1 \forall n$  and  $\liminf_{n \rightarrow \infty} \gamma_i^n > \gamma_i^0 \forall i$  imply again  $\gamma_o^n(q_n) \rightarrow \gamma_o^0(q_o)$ .

Let  $g_k^n: (0,1) \rightarrow FA$ :  $g_k^n(q) = \gamma_o^n(q)v_o^n + \sum_{i=1}^k \gamma_i^n(q)v_i$ . Since we have shown that  $v_o^n$  is norm convergent to  $v_o^0$  (or  $\gamma_o^0 = 0$ ) and that the  $\gamma_i^n(q)$  are equicontinuous and converging to  $\gamma_i^0(q)$ , it follows that  $\forall k$  the  $g_k^n$  form a sequence of continuous maps uniformly convergent on every compact set. If  $g^n = g_\infty^n$ , the relation

$\limsup_{k \rightarrow \infty} \sup_n \sum_i \gamma_i^n(q) = 0$  yields then that also  $g^n$  is a sequence of continuous maps uniformly convergent on compact sets to  $g^0$ .

Therefore, if  $f^n$  converges to  $f^0$  at every point of continuity of  $f^0$ ,  $\int_0^1 g^n(q) df^n(q)$  will converge to  $\int_0^1 g^0(q) df^0(q)$ : we have shown that:

Proposition 3: At every point where  $\mu$  has infinitely many players, value  $(f(\mu))$  - as a mapping from  $bv'([0,1]) \times FA_+^1$  to  $FA$  - is jointly continuous in  $f$  and  $\mu$ , when  $FA$  is endowed with the norm topology and  $bv'([0,1])$  is endowed with the strongest locally convex topology for which a sequence is convergent iff it has uniformly bounded variation and converges pointwise to the limit at every point of continuity of this (i.e., an Arens-topology, or bounded weak\*-topology).

Remark 2 (Regular games): Let  $v$  be a monotonic simple game with countably many players. Coalitions being points of  $\{0,1\}^\omega$ ,  $v$  is a  $\{0,1\}$ -valued monotonic function on  $\{0,1\}^\omega$ . Assume first  $v$  to be measurable for any product measure on  $\{0,1\}^\omega$  (in order for the extension to be defined - this assumption has to be made explicitly: indeed, using the continuum hypothesis, it is possible to construct such  $v$ 's such that the lower integral would be zero for the product of any sequences  $p_i$  with  $\limsup p_i < 1$  and the upper integral would be 1 whenever  $\liminf p_i > 0$ : there is little hope to be able to define a meaningful value for such things). We will also denote by  $v$  its extension to  $[0,1]^\omega$  defined by letting  $v(p_1, p_2, p_3, \dots)$  be the expectation of  $v$  under the corresponding product measure. We assume

$v$  to be continuous in the product topology in a uniform neighborhood of the diagonal - to exclude such obviously non regular games as:  $v(S) = 1$  iff  $\liminf_{n \rightarrow \infty} \{\text{proportion of players belonging to } S \text{ among the first } n \text{ players}\} > 1/2$ . Such a game is called regular (or non-singular, or proper) (cfr. Shapiro and Shapley [1971], at least for weighted majority games) if  $\sum_i \pi_i = 1$ , where  $\pi_i = P(i \text{ pivots})$ .

Remark now that  $\pi_i = \int_0^1 P(i \text{ pivots arriving at } t) dt = \int_0^1 [v(t, t, t, \dots, t, 1, t, \dots) - v(t, t, t, \dots, t, 0, t, \dots)] dt = \int_0^1 (\partial v / \partial p_i)(t, t, t, \dots) dt$  (this last formula because  $v$  is obviously multilinear in any finite number of  $p_i$ 's). The same multilinearity yields therefore that, for any sequence  $(\delta_i)$  with finitely many non zero terms,

$$\begin{aligned} \lim_{\tau \rightarrow 0} \int_0^1 \frac{v(t + \tau \delta_i) - v(t)}{\tau} dt &= \int_0^1 \lim_{\tau \rightarrow 0} \frac{v(t + \tau \delta_i) - v(t)}{\tau} dt \\ &= \int_0^1 \sum_i \delta_i \left( \frac{\partial v}{\partial p_i} \right)(t, t, \dots) dt = \sum_i \delta_i \pi_i . \end{aligned}$$

Therefore for any nonnegative sequence  $\delta_i$  ( $0 < \delta_i < 1$ ) if  $\delta_i^n$  denotes the same sequence with all but the first  $n$  terms set to zero we get

$$\begin{aligned} \liminf_{\tau \rightarrow 0} \int_0^1 \frac{v(t + \tau \delta_i^n) - v(t)}{\tau} dt &> \lim_{n \rightarrow \infty} \lim_{\tau \rightarrow 0} \frac{v(t + \tau \delta_i^n) - v(t)}{\tau} dt \\ &= \lim_{n \rightarrow \infty} \sum_i \delta_i^n \pi_i = \sum_i \delta_i \pi_i . \end{aligned}$$

By an argument we already made before this implies, when applied

also to the sequence  $1 - \delta_i$ , that, if  $\sum \pi_i = 1$ , then

$$\lim_{\tau \rightarrow 0} \int_0^1 \frac{v(t \cdot 1 + \tau \delta) - v(t \cdot 1)}{\tau} dt = \sum \delta_i \pi_i ,$$

i.e.,  $v \in Q$  with  $\psi(v) = (\pi_i)$ .

Thus, if  $v$  is regular (i.e.  $\sum \pi_i = 1$ ), then  $v \in Q$ . Conversely, if  $v \in Q$ , then  $\psi(v)$  is the limit of a sequence of continuous functions on  $[-1,1]^\infty$ , so is continuous at at least one point of this space, which implies  $\psi(v) \in \ell_1$  (this argument is essentially similar to an argument we already made in "Values and Derivatives"):  $\psi(v)$  is some summable sequence  $\psi_i$ . Since by our above argument one has  $\psi_i = \pi_i$ , and since efficiency yields  $\sum \psi_i = 1$  we get  $\sum \pi_i = 1$ : a monotonic, simple game with countably many players is in  $Q$  if and only if regular.

## SECTION 2

In Section 1 we have shown how to reduce the problem of defining a value to the problem of defining a positive, symmetric linear operator (of norm 1)  $\psi$  to FA from a (closed, symmetric) space  $V$  of functions  $v: B(I,C) \rightarrow R$  that satisfy  $v(a + bx) = av(1) + bv(x)$   $\forall a, b \in R \forall x \in B(I,C)$ .

We have also seen that, for such functions  $v$ , one has

$$|v(v)(x, x')| \leq \|x' - x\| \cdot \|v\| , \quad \forall x, x' \in B(I,C) .$$

Therefore, if we let  $D_x^\lambda(\tilde{x}) = [v(\tilde{x} + \lambda x) + v(\tilde{x} - \lambda x)]/2$ , we get

$$v(D_x^\lambda)[\tilde{x}, \tilde{x}'] \leq \|\tilde{x}' - \tilde{x}\| \cdot \|v\|$$

and  $D_{a+b\tilde{x}}^\lambda(c + d\tilde{x}) = cv(1) + d D_x^{|b\lambda/d|}(\tilde{x})$  (in particular  $D_x^\lambda(1) = v(1)$ ).

Let  $D_x^\lambda(\tilde{x})$  stand for  $\lim_{\lambda \rightarrow \pm\infty} D_x^\lambda(\tilde{x})$  (if this limit does not necessarily exist, use any Banach limit; remark that  $D_x^\lambda(\tilde{x})$  is necessarily an even function of  $\lambda$ ). We get then

$$v(D_x^\lambda)[\tilde{x}, \tilde{x}'] \leq \|\tilde{x}' - \tilde{x}\| \cdot \|v\|$$

and  $D_{a+b\tilde{x}}(c + d\tilde{x}) = cv(1) + dD_x^\lambda(\tilde{x})$ .

Thus,  $\forall x, D_x(\cdot)$  is linear on every plane containing the constants, and satisfies  $D_x(1) = v(1)$  and  $\|D_x\| \leq \|v\|$ .

In addition, the mapping  $x \mapsto D_x$  is constant on every plane containing the constants, and the mapping  $v \mapsto D_x$  is linear, positive and of norm 1.

$D_x(\tilde{x})$  is the (two-sided) derivative of  $v$  at  $x$  in the direction of  $\tilde{x}$ :  $\lim_{\tau \rightarrow 0} (v(x + \tau\tilde{x}) - v(x - \tau\tilde{x}))/2\tau$ .

We think of the typical situation where  $D_x$  would already be in FA for "almost every"  $x$ : for an average of the  $D_x$  then to be a value, one only has to make sure to get the symmetry. the average should be computed with a (finitely additive) probability distribution of  $B(I, C)$  that is invariant under all automorphisms of  $(I, C)$  (or of  $C$ ).

The averaging should be well defined whenever  $v$  is a function of a vector measure, so for any vector measure  $\mu = (\mu_1 \dots \mu_n)$  and for any Borel set  $B$  in  $R^n$ ,  $\mu^{-1}(B) = \{x | \mu(x) \in B\}$  should be measurable:

this class of sets in  $B(I, \mathcal{C})$  is the algebra of cylinder sets. Thus we look for a "cylinder probability" on  $B(I, \mathcal{C})$ , i.e., a finitely additive measure  $P$  on the cylinder sets, such that, for any vector measure  $\mu = (\mu_1 \dots \mu_n)$  the induced measure  $p \circ \mu^{-1}$  is a (countably additive) probability on the Borel sets of  $\mathbb{R}^n$ .

Recall that any cylinder probability on  $B(I, \mathcal{C})$  is uniquely characterized by its Fourier transform, a function on the dual defined by

$$F(\mu) = E \exp i\langle \mu, \chi \rangle .$$

In the next theorem we use the classical concept of invariance (i.e., under all automorphisms of  $(I, \mathcal{C})$ ); accordingly  $(I, \mathcal{C})$  is here required to be a standard Borel space (i.e., isomorphic to  $[0,1]$  with the Borel sets) and the duality used is that of  $B(I, \mathcal{C})$  with the space  $NA$  non atomic, countably additive measures on  $(I, \mathcal{C})$ .

Theorem 1: The extreme points of the set of invariant cylinder probabilities on  $B(I, \mathcal{C})$  have Fourier transforms  $F_{m,\sigma}(\mu) = \exp(im\mu(1) - \sigma\|\mu\|)$  where  $m \in \mathbb{R}$ ,  $\sigma > 0$ . More precisely, the formula  $E \exp i\langle \mu, \chi \rangle = \int_{\mathbb{R} \times \mathbb{R}_+} F_{m,\sigma}(\mu) dP(m, \sigma)$  establishes a one to one correspondence between invariant cylinder measures<sup>2/</sup> and (countably additive) measures  $P$  over  $\mathbb{R} \times \mathbb{R}_+$ . This correspondence is a positive, linear, convolution preserving isometry.

Proof: Consider first cylinder probabilities. Let  $\mu_i$  denote a sequence of mutually singular non atomic probabilities. There exists a

partition of  $(I, \mathcal{C})$  into a sequence of Borel sets  $B_i$ , such that  $\mu_i$  is carried by  $B_i$  - which has therefore the power of the continuum. Thus, for any permutation  $\pi$  of the integers, there exists an automorphism  $\theta_\pi$  of  $(I, \mathcal{C})$  such that  $\theta_\pi$  maps the sequence  $(\mu_i)_{i=1}^\infty$  to the sequence  $(\mu_{\pi(i)})_{i=1}^\infty$ .

The sequence  $\mu_i$  maps  $B(I, \mathcal{C})$  to  $\mathbb{R}^\infty$ , and the cylinder measure induces therefore a consistent system of probabilities on the Borel sets of the  $\prod_{i=1}^n (R)$ , and thus a (countably additive) probability  $Q$  on the Borel sets of  $\mathbb{R}^\infty$ . The invariance of the cylinder measure under  $\theta_\pi$  implies then the invariance of this probability under any permutation  $\pi$ : the coordinates of  $\mathbb{R}^\infty$  are exchangeable under  $Q$ .

Thus, by de Finetti's theorem, if we denote by  $A$  the asymptotic  $\sigma$ -field on  $\mathbb{R}^\infty$ , the random variables  $\mu_i$  are i.i.d. conditionally to  $A$ , say with distribution  $F$ . The mapping from  $\mathbb{R}^\infty$  to the set  $M(R)$  of probabilities on  $R$  that maps any sequence to its distribution (if this exists - which has  $Q$ -probability one by the Glivenko-Cantelli and de Finetti theorems) is  $A$ -measurable, so  $Q$  induces a probability  $P$  on  $M(R)$ , such that  $Q$  is the distribution of a sequence  $F^{-1}(x_i)$  where  $F$  is selected according to  $P$  and the  $x_i$  are selected, independently of  $F$  and of each other, uniformly on  $(0,1)$ . It follows in particular that any subsequence of the  $\mu_i$ 's would induce the same probability  $P$  on  $M(R)$ .

Let now  $\mu'_i$  be another such sequence; then there exists an uncountable Borel set  $B$  in  $(I, \mathcal{C})$  which is negligible for all  $\mu_i$ 's and  $\mu'_i$ 's: one can construct on  $B$  a third such sequence  $\mu''_i$ .

When the  $\mu_i$ 's and the  $\tilde{\mu}_i$ 's are arranged in sequence, they fulfill the requirements set out at the start of the proof, so the probability  $P$  on  $M(R)$  induced by the two subsequences  $\mu_i$  and  $\tilde{\mu}_i$  is the same. The same would apply to the two sequences  $\mu'_i$  and  $\tilde{\mu}_i$ , so it follows that  $P$  is independent of the particular sequence  $\mu_i$  chosen, but depends only on the cylinder measure.

Since  $\bar{\mu}_n = (1/n) \sum_{i=1}^n \mu_i$  is such that the sequence  $(\bar{\mu}_n, \bar{\mu}_{n+1}, \dots)$  satisfies our requirements, and has the same asymptotic  $\sigma$ -field as the original sequence, it follows that, for  $P$ -almost every  $F$ ,  $\mu_1 \dots \mu_n$  are independently  $F$ -distributed and  $\bar{\mu}_n = (1/n) \sum_{i \leq n} \mu_i$  is also  $F$  distributed. Thus  $P$ -almost every  $F$  is such that, for all  $n$ , the average of  $n$  independent  $F$ -distributed random variables is  $F$ -distributed, i.e.,  $F$  is strictly stable of index 1. For symmetric random variables, this is equivalent to say  $F$  is a Cauchy distribution.

If we parameterize the Cauchy distribution by their location and scale parameters  $m$  and  $\sigma$ ,  $P$  becomes a probability distribution on  $R \times R_+$  such that, for any sequence  $\mu_i$  of mutually singular non atomic measures, the sequence  $\mu_i(x)$  is distributed as the average under  $P(dm, d\sigma)$  of the distribution of  $(\|\mu_i^+\| \cdot X_i - \|\mu_i^-\| \cdot Y_i)_{i=1}^\infty$ , where the  $X_i$  and  $Y_i$  are all independently distributed as  $m + \sigma U$ , where  $U$  is a standard Cauchy random variable.

Thus  $\|\mu_i^+\| \cdot X_i - \|\mu_i^-\| \cdot Y_i$  is distributed like  $m \cdot \mu_i(1) + \sigma \cdot \|\mu_i\| \cdot U_i$ , where  $U_i$  is a standard Cauchy random variable.

In particular,  $E[\exp(i\langle \mu, x \rangle) | m, \sigma] = \exp[-\sigma \|\mu\| + im \langle \mu, 1 \rangle]$

$$\forall \mu \in NA, \text{ and } E \exp(i\langle \mu, \chi \rangle) = \int_{R^+ \times R} \exp(-\sigma \|\mu\| + im\langle \mu, l \rangle) dP(m, \sigma).$$

It is clear from the above proof - or from the last formula and the uniqueness theorems for Fourier and Laplace transforms - that for any cylinder measure there can exist only one  $P$  such that the above formula holds.

Let us now show that for any such  $P$  there exists a (unique) cylinder measure with that Fourier transform.

Uniqueness follows immediately from the fact that distributions over finite dimensional spaces are uniquely characterized by their Fourier transform, and from the fact that cylinder sets are all finite dimensional sets. To show existence, recall that Bochner's theorem characterizes the characteristic functions on  $\overset{\sim}{R^n}$  as the positive definite functions  $\psi$  that are continuous at zero with  $\psi(0) = 1$ . This immediately extends itself to Fourier transforms of cylinder probabilities (when continuity at zero is interpreted as continuity at zero of the restrictions to all finite dimensional subspaces of the dual).

Indeed every inequality for positive definiteness involves only finitely many points in the dual, so the condition is still necessary, and if it holds, we get by Bochner's theorem a consistent system of probability distributions on all finite dimensional quotient spaces of  $B(I, C)$ , i.e., a cylinder probability.

Now our formula obviously has value 1 at zero, and is continuous there by the dominated convergence theorem. Thus we only have to show that it is positive definite. For this it is sufficient to show that for every  $(m, \sigma)$  the function  $\exp(-\sigma \|\mu\| + im\langle \mu, l \rangle)$  is positive definite, the inequalities being linear.

To show this, it is sufficient to show that this function is the pointwise limit of a set of positive definite functions  $\psi_\alpha$ , since the inequalities each involve only a finite number of points in the dual and are weak inequalities.

For every Borel partition  $\alpha = (B_1^\alpha \dots B_n^\alpha)$  of  $(I, \mathcal{C})$ , let  $X_1(\omega) \dots X_n(\omega)$  be independent Cauchy random variables with parameters  $m$  and  $\sigma$ , and let  $f(\omega) \in B(I, \mathcal{C})$  have value  $X_j(\omega)$  on  $B_j^\alpha$ :  $f(\omega)$  is a random variable with values in (a finite dimensional subspace of)  $B(I, \mathcal{C})$ , thus by Bochner's theorem its characteristic function  $\psi_\alpha$  will be positive definite. We have  $\psi_\alpha(\mu) = E \exp(i\langle \mu, \chi \rangle) = E \exp(i\langle \mu, f(\omega) \rangle) = E \exp i \sum_{j=1}^n \mu(B_j^\alpha) X_j(\omega)$ .

Now  $\sum_{j=1}^n \mu(B_j^\alpha) X_j(\omega)$  is Cauchy with parameters  $m \sum_j \mu(B_j^\alpha)$  and  $\sigma \sqrt{\sum_j \mu(B_j^\alpha)}$ , i.e.,  $m \cdot \langle \mu, 1 \rangle$  and  $\sigma \|\mu\|_\alpha$ , where  $\|\mu\|_\alpha$  is the norm of the restriction of  $\mu$  to the ( $\sigma$ -field generated by the) partition  $\alpha$ .

Thus  $\psi_\alpha(\mu) = \exp[-\sigma \|\mu\|_\alpha + im\langle \mu, 1 \rangle]$  is positive definite, and obviously  $\|\mu\|_\alpha \rightarrow \|\mu\|$  when  $\alpha$  ranges over the increasing net of all partitions.

This proves that  $\psi$  is positive definite, and thereby establishes the one to one character of this correspondence, when restricted to probabilities on both sides. (Obviously the cylinder probability has to be invariant, since its Fourier transform is so.)

It is now clear that, for any bounded measure  $P$ , there exists a corresponding invariant cylinder measure: let  $P = \alpha P_1 - \beta P_2$ , where  $P_1$  and  $P_2$  are two probabilities,  $\alpha > 0$ ,  $\beta > 0$ , and use  $\alpha Q_1 - \beta Q_2$  as invariant cylinder measure, where  $Q_i$  is the cylinder probability

corresponding to  $P_i$ . Furthermore this cylinder measure is unique - if there were two of them, their difference would be a cylinder measure with zero Fourier transform, so the positive and negative parts of this difference would be two different positive cylinder measures with the same Fourier transform, and in particular with the same total mass (value of the Fourier transform at zero), so that by normalizing one would obtain two different cylinder probabilities with the same Fourier transform, in contradiction with what we have seen above.

We have just used the fact that the positive part  $\lambda^+$  of a (bounded) cylinder measure  $\lambda$  is still a cylinder measure. Indeed, if  $A$  denotes the algebra of cylinder sets,  $\lambda^+$  is defined by  $\lambda^+(A) = \sup_{B \in A} \lambda(A \cap B) \quad \forall A \in A$ . One sees immediately that  $\lambda^+$  is finitely additive, positive and bounded on  $A$ , with  $\lambda^+ > \lambda$ . To show that  $\lambda^+$  is still a cylinder measure, let  $A_\phi = \{\phi^{-1}(B) | B \text{ Borel set in } \mathbb{R}^n\}$  for  $\phi$  ranging over all finite subsets  $\{\phi_1 \dots \phi_n\}$  of  $\mathbb{N}^A$ . Then  $\lambda^+ = \sup_\phi \lambda_\phi^+$ , with  $\lambda_\phi^+(A) = \sup_{B \in A_\phi} \lambda(A \cap B)$ . It is thus sufficient to show that,  $\forall \phi_0, \forall \phi: \phi > \phi_0, \lambda_\phi^+$  is countably additive on  $A_{\phi_0}$  - (the supremum of a bounded, increasing net of countably additive measures is still) or that,  $\forall \phi, \lambda_\phi^+$  is countably additive on  $A_\phi$ : this is the Hahn decomposition theorem for countably additive measures.

Obviously, if further  $\lambda$  was invariant,  $\lambda^+$  will also be: therefore, we can, in the same way as above for  $P$ , construct for any invariant cylinder measure  $\lambda$  a corresponding measure  $P$  on  $\mathbb{R} \times \mathbb{R}_+$ . Again, this  $P$  is unique, otherwise one could construct, as above, two different probabilities  $P_1$  and  $P_2$  with the same value of

the integral  $\int F_{m,\sigma}(\mu) dP(m,\sigma)$ , contradicting our previous result for probabilities.

Thus the bijectivity of the correspondence is established. Its positivity was already established before, when dealing with probabilities, and its linearity is now immediately obvious from the bijectivity - an integral is a linear function of the underlying measures. Being positive and linear, it is an isometry because it maps both ways probabilities to probabilities.

The assertion about extreme points is now immediate, so there only remains to establish the preservation of convolution.

Since a linear mapping from  $R^n$  to  $R^k$  maps the convolution of two measures to the convolution of their images, it is clear that the convolution of two cylinder measures is a well defined cylinder measure, with the Fourier transform of the convolution being the product of the Fourier transforms of the individual measures. In particular, if the two measures were invariant, the convolution will still be. Similarly one checks immediately that the integral in the right hand side under the convolution of two measures  $P_1$  and  $P_2$  is the product of the corresponding integrals. This finishes the proof. Q.E.D.

Denote by  $Q$  the closed, symmetric space generated by  $FA$  and all functions  $v$  satisfying  $v(a + bx) = av(1) + bv(x)$ ,  $\|v\| < \infty$  that are of the form  $v(x) = f(\mu(x))$ , where  $\mu$  is a vector measure in  $NA$ .

Theorem 2: Let  $v \in Q$ , and let  $P$  be any invariant cylinder measure of total mass 1 on  $B(I,C)$  which is nondegenerate, i.e., the

subspace of constant functions has probability zero, or: Prob ( $\sigma = 0$ ) = 0.

Then  $D_{\tilde{X}}(\tilde{x})$  exists, for every  $\tilde{x}$ , for  $P$ -almost every  $x$  (i.e., the difference  $\sup_{\lambda > \lambda_0} D_x^\lambda(\tilde{x}) - \inf_{\lambda > \lambda_0} D_x^\lambda(\tilde{x})$  converges to zero in  $L_1(dP(x))$  when  $\lambda_0 \rightarrow \infty$ ) and is, as well as any  $D_x^\lambda(\tilde{x})$ ,  $P$ -integrable in  $X$ , and  $\phi_v(\tilde{x}) = \int D_{\tilde{X}}(\tilde{x}) dP(x) = \lim_{\lambda \rightarrow \infty} \int D_x^\lambda(\tilde{x}) dP(x)$  is independent of the particular invariant  $P$  chosen.

Further  $\phi_v \in FA$ , so that the mapping  $v \mapsto \phi_v$  is positive, linear, symmetric, of norm 1, and satisfying  $\phi_v(1) = v(1)$ :  $\phi: v \mapsto \phi_v$  is a value on  $Q$ .

Proof: Since the mapping  $v \mapsto D_x^\lambda$  is positive, linear, of norm 1 and satisfies  $D_x^\lambda(1) = v(1)$ , the last sentence of the statement will follow from the others provided we prove the additivity of  $\phi_v$ .

It also follows that it is sufficient to prove the statement on the generators of the space, since a uniform limit of  $P$ -integrable functions is  $P$ -integrable, with the integral being continuous along the sequence.

Finally, since  $D_x^\lambda$  acts as the identity on  $FA$ , and since constant functions are  $P$ -integrable, it is sufficient to consider the generators of the form  $v = f(\mu)$ , with  $\mu = (\mu_1 \dots \mu_n)$  a vector measure in  $NA$ .

Also, since, by Theorem 1,  $P$  can be written as  $\alpha P_1 - \beta P_2$ , where the  $P_i$  are invariant cylinder probabilities and  $\alpha - \beta = 1$ , it is sufficient to consider the case where  $P$  is a cylinder probability.

There is no loss in assuming that  $\mu$  has full dimensional range - otherwise one of the components of  $\mu$  is a linear combination of the

others, so  $v$  can be written only as a function of the other components.

Denote by  $B_\mu$  the image under  $\mu$  of the unit ball of  $B(I, C)$ , i.e.,  $B_\mu = \{f(\text{Range of } \mu) - \mu(1)\}$ .  $B_\mu$  being compact, convex, symmetric around zero, and full dimensional, it is a neighborhood of zero (by the absorption theorem say). The relation  $v(a + bx) = av(1) + bv(x)$  implies now  $f(a \cdot e + bx) = af(e) + bf(x)$ ,  $\forall x \in R^n$ ,  $a, b \in R$ , where  $e = \mu(1) \in R^n$ .

Finally, the relation  $V(v)[x, x'] \leq \|x' - x\| \circ \|v\|$  implies that, if  $x = \mu(\tilde{x})$ , and  $y - x \in \delta B_\mu$ , then  $\exists \bar{x}$  with  $\|\bar{x} - \tilde{x}\| \leq \delta$  and  $y = \mu(\bar{x})$  so that

$$\begin{aligned}|f(y) - f(x)| &= |v(\bar{x}) - v(\tilde{x})| \leq |v(\bar{x}) - v(\tilde{x} - \delta \cdot 1)| + |v(\tilde{x} - \delta \cdot 1) - v(\tilde{x})| \\&\leq V(v)[\tilde{x} - \delta \cdot 1, \tilde{x} + \delta \cdot 1] + V(v)[\tilde{x} - \delta \cdot 1, \tilde{x}] \leq 3\delta\|v\|.\end{aligned}$$

$B_\mu$  being a neighborhood of zero,  $\exists \epsilon > 0$ :  $\|x\| < \epsilon \Rightarrow x \in B_\mu$ , and thus we have shown that  $\|y - x\| < \epsilon\delta \Rightarrow |f(y) - f(x)| < 3\delta\|v\|$  for all  $\delta$ ,  $y$  and  $x$ : thus  $|f(y) - f(x)| < (3\|v\|)/\epsilon \circ \|y - x\|$ :  $f$  is Lipschitz.

Conversely if  $f$  is Lipschitz it follows immediately that  $\|v\| < \infty$ , so our assumptions reduce simply to  $v = f(\mu)$ , where  $\mu$  is a vector in NA with full n-dimensional range and where  $f$  is Lipschitz satisfying  $f(a \cdot e + bx) = af(e) + bf(x)$  where  $e = \mu(1)$ ,  $\forall x \in R^n$ ,  $\forall a, b \in R$ .

We have  $D_x^{\lambda}(\tilde{x}) = [f(x + \lambda y) + f(x - \lambda y)]/2$ , where  $x = \mu(\tilde{x})$ ,  $y = \mu(\tilde{y})$  or  $D_x^{-1}(\tilde{x}) = [f(y + \tau x) - f(y - \tau x)]/(2\tau)$ .

Remark that,  $f$  being Lipschitz, the limit (for  $\tau \rightarrow 0$ ) will, for

each  $x$ , exist  $\lambda$ -almost everywhere in  $y$ ,  $\lambda$  being Lebesgue measure. This follows from Lebesgue's a.e. differentiability theorem. Indeed, if  $x$  is zero, there is nothing to prove, otherwise  $x$  can be taken as the first basis vector in  $R^n$ : for any  $z_2 \dots z_n$ ,  $f(z, z_2, \dots, z_n)$  is a Lipschitz function of  $z$ , so the first partial derivative exists for almost  $z$ , by Lebesgue's theorem. Since  $f$  is Lipschitz on  $R^n$ , the set of points where the first partial derivative does not exist is a Borel set, and therefore this set of points has Lebesgue measure zero by Fubini's theorem.

The probability induced by  $P$  on  $R^n$  has characteristic function

$$\psi_\mu(t) = E \exp i\langle t, x \rangle = E \exp i\langle t, \mu(x) \rangle = E \exp i(\langle t, \mu \rangle(x)) = \int_{R^n} \exp[-\sigma \|t, \mu\|] + i\text{im}(\langle t, \mu \rangle(1)) dP(m, \sigma) \quad \text{for some probability } P.$$

Now  $\|t, \mu\| = \sup_{\|x\| \leq 1} \langle t, \mu(x) \rangle = \sup_{x \in B_\mu} \langle t, x \rangle - N_\mu(t)$  where  $N_\mu$  is the norm on the dual generated by the ball  $B_\mu$ .

And  $\langle t, \mu \rangle(1) = \langle t, e \rangle$ . Thus

$$\psi_\mu(t) = \int_{R^n} \exp[-\sigma N_\mu(t) + i\text{im}\langle t, e \rangle] dP(m, \sigma).$$

Now obviously, for any given  $m$  and  $\sigma (>0)$ ,  $\exp[-\sigma N_\mu(t) + i\text{im}\langle t, e \rangle]$  is Lebesgue integrable in  $t$ , so the corresponding probability distribution has, by the Fourier inversion theorem, a density with respect to Lebesgue measure. Since the conditional distribution on  $R^n$  given  $m$  and  $\sigma$  is absolutely continuous, the unconditional distribution is also certainly so.

Thus we may conclude that, for any invariant  $P$ , and for any

$\tilde{x}$ , the limit  $D_x(\tilde{x})$  will exist for  $P$ -almost every  $x$ .

Thus, for any  $P$ , and any  $x$ ,  $[f(y + \tau x) - f(y - \tau x)]/(2\tau)$  is uniformly bounded ( $f$  being Lipschitz) and converges  $P$ -a.e. to its limit: by the dominated convergence theorem, the limit is  $P$ -integrable and the limit of the integrals is the integral of the limit function  $\psi_x(y)$ .

Now  $f(a \cdot e + by) = af(e) + bf(y)$  yields  $\psi_x(a \cdot e + by) = \lim_{\lambda \rightarrow \infty} [f(x + \lambda(a \cdot e + b \cdot y)) + f(x - \lambda(a \cdot e + b \cdot y))]/2 = \lim_{\lambda \rightarrow \infty} [f(x + \lambda by) + f(x - \lambda by)]/2 = \psi_x(y)$  if  $b \neq 0$  (and  $= f(x)$  if  $b = 0$ ).

Let  $Z$  denote a random variable having characteristic function  $\exp[-N_\mu(t)]$ . Then  $m \cdot e + \sigma Z$  where  $(m, \sigma)$  is selected, independently of  $Z$ , according to  $\tilde{P}(m, \sigma)$ , has the correct characteristic function  $\int_{R \times R} \exp[-\sigma N_\mu(t) + im \langle t, e \rangle] d\tilde{P}(m, \sigma)$ . Thus  $\int \psi_x(y) d\tilde{P}(y) = E[\psi_x(m \cdot e + \sigma Z)] = E\psi_x(Z)$  since  $\tilde{P}(\sigma = 0) = 0$ : the integral of the limit - which is the limit of the integrals - does not depend on the choice of  $\tilde{P}$ , i.e. on the choice of a particular invariant cylinder probability.

There only remains to establish the additivity, i.e., that  $E\psi_x(Z)$  is a linear function of  $x$ .

Let  $f_\epsilon(x) = f(x) \exp(-\epsilon \|x\|^2)$  ( $\epsilon > 0$ ) (here  $\| \cdot \|$  is the Euclidean norm).

We want to show that  $f_\epsilon$  are uniformly Lipschitz (i.e., with a Lipschitz constant independent of  $\epsilon$ ).

Since  $f$  is Lipschitz, each of them is obviously locally Lipschitz, so by the above mentioned theorem of Lebesgue, it will be sufficient to show that the directional derivatives of the  $f_\epsilon$  are uniformly bounded whenever they exist.

By choosing appropriate axes, we can assume our directional derivative is in the direction of the  $x_1$  axis.

We have

$$\frac{\partial f_\epsilon}{\partial x_1} = \left( \frac{\partial f}{\partial x_1} \right) \exp[-\epsilon \|x\|^2] - 2\epsilon x_1 f \exp[-\epsilon \|x\|^2].$$

If  $K$  is the Lipschitz constant of  $f$ , then  $|\partial f/\partial x_1| \leq K$  and  $|f| \leq K \|x\|$  - bounding also  $|x_1|$  by  $\|x\|$ , we get

$$|\frac{\partial f_\epsilon}{\partial x_1}| \leq K \exp[-\epsilon \|x\|^2] + 2K(\epsilon \|x\|^2) \exp[-(\epsilon \|x\|^2)] \leq 2K$$

$$\text{since } e^{-z} + 2ze^{-z} \leq 2.$$

Thus the  $f_\epsilon$  have uniformly the Lipschitz constant  $2K$ . Further the formula shows that, whenever the directional derivative of  $f$  exists, the corresponding directional derivatives of the  $f_\epsilon$  will also exist and converge to that of  $f$  when  $\epsilon \rightarrow 0$ .

Thus  $\psi_x(y) = \lim_{\epsilon \rightarrow 0} \lim_{\tau \rightarrow 0} [f_\epsilon(y + \tau x) - f_\epsilon(y - \tau x)]/(2\tau)$ , all functions involved being  $\leq 2K\|x\|$  in absolute value. Thus, by the dominated convergence theorem,

$$E\psi_x(z) = \lim_{\epsilon \rightarrow 0} \lim_{\tau \rightarrow 0} \frac{1}{2\tau} \int [f_\epsilon(z + \tau x) - f_\epsilon(z - \tau x)] g_\mu(z) dz$$

where  $g_\mu$  is the density of  $Z$  (which we have already shown to exist).

But since  $f_\epsilon$  is a bounded function, it is integrable, so

$$\begin{aligned}\int |f_\epsilon(z + \tau x) - f_\epsilon(z - \tau x)| g_\mu(z) dz &= \int f_\epsilon(z) g_\mu(z - \tau x) dz - \int f_\epsilon(z) g_\mu(z + \tau x) dz \\ &= \int f_\epsilon(z) [g_\mu(z - \tau x) - g_\mu(z + \tau x)] dz.\end{aligned}$$

Now  $g_\mu$  has characteristic function  $\exp[-N_\mu(t)]$ , and  $|t| \exp[-N_\mu(t)]$  is integrable for Lebesgue measure. Therefore, by the Riemann-Lebesgue theorem,  $g_\mu$  is continuously differentiable with its gradient going to zero at infinity. In particular the  $[g_\mu(z - \tau x) - g_\mu(z + \tau x)]/2\tau$  are uniformly bounded and converge pointwise to  $\langle -(\nabla g_\mu)(z), x \rangle$ , where  $(\nabla g_\mu)(z)$  is the gradient of  $g_\mu$  at  $z$ .

Since  $f_\epsilon(z)$  is integrable, it follows (dominated convergence) that  $\lim_{\tau \rightarrow 0} (1/2\tau) \int f_\epsilon(z) [g_\mu(z - \tau x) - g_\mu(z + \tau x)] dz = \int f_\epsilon(z) \langle -(\nabla g_\mu)(z), x \rangle dz$  and thus

$$\begin{aligned}E\psi_x(z) &= \lim_{\epsilon \rightarrow 0} \int e^{-\epsilon \|z\|^2} f(z) \langle -(\nabla g_\mu)(z), x \rangle dz \\ &= \langle x, \lim_{\epsilon \rightarrow 0} \int e^{-\epsilon \|z\|^2} f(z) (\nabla g_\mu)(z) dz \rangle\end{aligned}$$

which is linear in  $x$  (the limit being some form of Cauchy principal value of  $\int f(z) (\nabla g_\mu)(z) dz$ ).

This finishes the proof. Q.E.D.

Remarks:

- 1) One can use the same formula ( $\exp[-\|\mu\|]$ ) to define the Fourier transform on the whole of FA, thereby defining a cylinder

probability on  $B(I, \mathcal{C})$  when cylinder sets are defined as inverse images of Borel sets by any vector  $(\mu_1 \dots \mu_n)$  in FA. This cylinder probability would obviously be invariant under all automorphisms of the lattice  $\mathcal{C}$ .

Theorem 2 remains then valid when  $\mu$  in the definition of  $Q$  is allowed to be any vector in FA - provided one interprets "invariant P" as " $\Rightarrow$  having the Fourier transform prescribed by Theorem 1."

It might be that this formula could be justified by some type of uniqueness argument on the space of non atomic elements of FA - using maybe a weaker concept of automorphism. But certainly for the atomic part no such argument could be hoped for.

However, as our analysis of regular games with countably many players at the end of Section 1 may indicate, it could be that in general the "atomic part of the game" is already essentially linearized by the first derivative operation, so that the end result would anyway be canonical. This certainly deserves further study.

2) Define for any vector measure  $\mu$ ,  $N_\mu(t)$  as  $\sup_{x \in B_\mu} \langle t, x \rangle = \| \langle t, \mu \rangle \| = \int |\langle t, x \rangle| d\nu(x)$ , where  $\nu$  is the distribution, under a common dominating measure  $\mu_0$ , of the Radon Nikodym derivatives  $f = (f_1 \dots f_n)$  of  $\mu = (\mu_1 \dots \mu_n)$  w.r.t.  $\mu_0$ . For any norm  $n$ , one could replace  $f$  by  $f' = f/n(f)$ , and  $d\mu_0$  by  $d\mu'_0 = n(f)d\mu_0$  to normalize  $\nu$  on the  $n$ -unit sphere, say, for a canonical choice,  $\nu$  could be carried by the boundary of  $B_\mu$ .

Our proof then shows that, in this case,  $\exp(-N_\mu(t))$  is positive definite. Conversely however, if the support function  $N_\mu(t) = \sup_{x \in B_\mu} \langle t, x \rangle$

of some compact, convex, symmetric set  $B_\mu$  is such that  $\exp[-N_\mu(t)]$  is positive definite, then, since  $N_\mu$  is positively homogeneous of degree one,  $\exp[-N_\mu(t)]$  is the characteristic function of a strictly stable distribution of index one, and has therefore as classical Levy representation  $\exp[-\int |\langle t, x \rangle| dv(x)]$ , where  $v$  is the normalization of the Levy measure of the process on some sphere - say on the boundary of  $B_\mu$ : there exists a positive measure  $v$  on the boundary of  $B_\mu$  such that  $N_\mu(t) = \int |\langle t, x \rangle| dv(x)$ . If now we define  $\mu$  by  $d\mu_i = x_i dv$ , where  $x_i$  is the  $i^{\text{th}}$  coordinate mapping, we get immediately  $N_\mu(t) = \| \langle t, \mu \rangle \|$ :  $B_\mu$  is indeed the ball corresponding to the vector measure  $\mu$ .

This interpretation in terms of the Levy measure allows us therefore to view the random perturbation around the diagonal as the sum of a very large number of independent contributions - those of the players preceding the given player in a random order - the direction of each being according to the distribution of Radon Nikodym derivatives of the given measure. This type of interpretation will be pursued much further in a subsequent paper.

3) A large number of definitions of "spaces on which there is a value" are possible in view of what precedes - depending among others on the exact order in which the various limiting operations and averaging operations are to be done, on how much "a.e." is put into the definitions, etc.

We prefer to leave this matter to the taste of the reader, as long as no theorems are available that would show clearly which option is to be

preferred. For a foretaste, the reader may want to look at Section 4.

### SECTION 3

In many applications of the above results, whether to majority games with several houses or to  $n$ -handed glove markets for instance, the function  $v$  of Section 2 will be of the form  $f(\mu_1 \dots \mu_n)$ , where  $\mu$  is a vector measure and  $f$  is piecewise linear. Elementary transformations reduce this to the case where  $\mu$  is a full dimensional vector of probability measures. Say  $f_i$  ( $i = 1 \dots k$ ) are the different linear functions appearing as pieces of  $f$  ( $i \neq j \Rightarrow f_i \neq f_j$ ). Then the set  $\{x | f_i(x) = f_j(x)\}$  being of lower dimension, has zero probability under the invariant measure of the last section (since this is absolutely continuous with respect to Lebesgue measure), so that we can neglect ties among the  $f_i$ 's. Then, for any order  $\prec$  on the indexes  $1 \dots k$ , the set  $\{x | \forall i, j, i \prec j \Rightarrow f_i(x) < f_j(x)\} = C(\prec)$  is an open convex cone - thus connected - where, by continuity,  $f$  is constantly equal to one of the  $f_i$ 's - say  $f_{i(\prec)}$ .

Thus, by the results of Section 2, the value of this game takes the form  $\sum_{\prec} P[C(\prec)] f_{i(\prec)}(\mu)$ .

So, to compute the value of such games, we have to compute the probability that  $f_{i_1}(\mu(x)) < f_{i_2}(\mu(x)) < \dots < f_{i_k}(\mu(x))$  - or, letting  $\Phi_j$  stand for the measure  $f_{i_j}(\mu)$ , the probability that  $\Phi_1 < \Phi_2 < \dots < \Phi_k$ , when  $\Phi$  is some vector measure. Remark also that the property  $f(t \cdot 1 + a \cdot x) = tf(1) + af(x)$  implies that, for all  $i$  needed to

represent  $f$  (i.e.,  $f = f_i$  on some open set), one has not only  $f_i$  linear and not merely affine, so the  $\Phi_j$  are indeed measures, but also  $f_i(1) = f(1)$ , so they also have the same total mass.

Thus, letting  $v_i = \Phi_{i+1} - \Phi_i$ , we have a vector measure with total mass zero, and we have to compute the probability that  $v(x)$  falls in the positive orthant.

If the  $v_i$ 's are not linearly independent, those inequalities determine a convex polyhedral cone in the space generated by  $v$ . This cone can be written as a finite union of convex simplicial cones (neglecting boundaries that have probability zero), and for each convex simplicial cone one can take its extreme rays as new coordinate axes, thus reverting to the case where the  $v_i$ 's are linearly independent.

This is the probability we are going to compute in this section:  $v = (v_1 \dots v_n)$  is a full dimensional vector measure with total mass zero, and we want  $P(v(x) \in R^n_+)$ .

Obviously, this probability does not depend on the particular invariant measure chosen, so we will use  $m = 0$ ,  $\sigma = 1$ .

Let us first recall that for any norm  $N$  on  $R^n$ , any point  $x \in R^n$  can be written in polar coordinates  $r = N(x)$  and  $s = x/r$ , and that Lebesgue measure  $dx_1 \dots dx_n = r^{n-1} dr d\sigma(s)$ , by definition of the surface measure  $d\sigma$  on the unit  $N$ -sphere. One gets the following "change of variables" formula: if  $\tau$  is any other such surface measure (i.e., originating from some other norm), then for any positive measurable function  $f$  on the unit  $N$ -sphere,

$$\int f(s) d\sigma(s) = \int f \left[ \frac{a}{N(a)} \right] \frac{d\tau(a)}{N^n(a)} .$$

From now on we denote shortly by  $N$  the support function  $N_v$  of  $B_v$ .

We observed in Section 2 that the characteristic function  $\exp[-N(\tau)]$  is integrable, so the Fourier inversion formula holds. Thus

$$P = P(v(x) \in R_+^n)$$

$$= \frac{1}{(2\pi)^n} \int_{R_n^+} dx_1 \dots dx_n \int [\exp(-N(y))] [\exp(-i\langle y, x \rangle)] dy_1 \dots dy_n$$

or, going to polar coordinates

$$\begin{aligned} P &= (2\pi)^{-n} \int_{R_n^+} (dx_1 \dots dx_n) \int \exp[-r(1 + i\langle s, x \rangle)] r^{n-1} dr d\sigma(s) \\ &= \frac{(n-1)!}{(2\pi)^n} \int_{x_i > 0} dx_1 \dots dx_n \int \frac{d\sigma(s)}{[1 + i\langle s, x \rangle]^n}. \end{aligned}$$

The inner integral being a density, it is positive, so we get from the monotone convergence theorem

$$P = \frac{(n-1)!}{(2\pi)^n} \lim_{M \rightarrow \infty} \int_{0 < x_i < M} dx_1 \dots dx_n \int \frac{d\sigma(s)}{[1 + i\langle s, x \rangle]^n}.$$

Now  $1/([1 + i\langle s, x \rangle]^n)$  is bounded (its absolute value being  $< 1$ ) and thus integrable on the product of any cube in  $x$  and the unit  $N$ -sphere. Using thus Fubini's theorem, we get

$$P = \frac{(n-1)!}{(2\pi)^n} \lim_{M \rightarrow \infty} \int d\sigma(s) \int_{0 < x_i < M} \frac{dx_1 \dots dx_n}{[1 + i\langle s, x \rangle]^n}.$$

$$\text{Let } \phi_n(c, s) = i^n (n-1)! \prod_{i=1}^n \int_{0 < x_i < M} (dx_1 \dots dx_n) / ([c + i\langle s, x \rangle]^n),$$

( $\operatorname{Re}(c) = 1$ ):  $\phi_n$  depends only on the first  $n$  coordinates of the sequence

$s_i$ . An elementary integration over  $x_n$  yields

$$\phi_n(c, s) = \sum_{\delta_n \in \{0, 1\}} (-1)^{\delta_n} \phi_{n-1}[c + i\delta_n s_n M, s]$$

and this formula still holds for  $n = 1$  if one sets  $\phi_0(c) = -\ln c$ .

One gets now immediately by induction that

$$\phi_n(c, s) = - \sum_{\delta \in \{0, 1\}^n} (-1)^{\sum_j \delta_j} \ln(c + iM \sum_j \delta_j s_j),$$

and thus

$$P = \frac{-1}{(2\pi i)^n} \lim_{M \rightarrow \infty} \int \left[ \sum_{\delta \in \{0, 1\}^n} (-1)^{\sum_j \delta_j} \ln(1 + iM \sum_j \delta_j s_j) \right] \frac{d\sigma(s)}{ds} .$$

Since  $d\sigma(s)$  is symmetric around zero, we can replace each

$$\frac{-1}{i^n} \frac{\ln(1 + iM \sum_j \delta_j s_j)}{ds_j}$$

by the average of its value at  $s$  and at  $(-s)$ , i.e., by its real part. We get thus

$$P = \frac{1}{(2\pi)^n} \lim_{M \rightarrow \infty} \int \left[ \sum_{\delta \in \{0, 1\}^n} (-1)^{\sum_j \delta_j} F_n(M \sum_j \delta_j s_j) \right] \frac{d\sigma(s)}{ds_j}$$

where

$$F_k(x) = -\frac{1}{2} \ln(1 + x^2), -\text{Arctan}(x), \frac{1}{2} \ln(1 + x^2), \text{Arctan}(x)$$

according as to  $k = 0, 1, 2$  or  $3 \pmod 4$ .

Here  $\text{Arctan}(x)$  denotes the inverse of the tangent function, with

values in  $(-\pi/2, \pi/2)$ .

Using now the change of variables formula, we can rewrite this as

$$P = \frac{1}{(2\pi)^n} \lim_{M \rightarrow \infty} \int \left[ \sum_{\delta \in \{0,1\}^n} (-1)^{\sum_j \delta_j} F_n\left(\frac{M}{N(s)} \sum_j \delta_j s_j\right) \right] \frac{d\tau(s)}{\pi s_j}$$

where  $\tau$  denotes the surface measure corresponding to an arbitrary norm

II. Henceforth we will use  $\|x\| = \sum_i |x_i|$ . For this norm, the unit sphere has  $2^n$  faces, each with  $\tau$ -area equal to  $1/(n-1)!$ .

Letting  $\Delta_n = \{s; s > 0, \sum_i s_i = 1\}$  we get, folding all faces back on  $\Delta_n$ ,

$$P = (2\pi)^{-n} \lim_{M \rightarrow \infty} \int \left[ \sum_{\substack{\delta \in \{0,1\}^n \\ \epsilon \in \{-1,1\}^n}} (-1)^{\sum_j \delta_j (\pi \epsilon_j)} F_n\left(\frac{M}{N(\epsilon \cdot s)} \sum_j \delta_j \epsilon_j s_j\right) \right] \frac{d\tau(s)}{\pi s_j} .$$

The next part of the computation is for  $n$  even.

$$\text{Let } \phi_\delta(M, s) = (1/2) \sum_{\epsilon \in \{-1,1\}^n} (\pi \epsilon_j) \ln\left(1 + \left[\frac{(M/N(\epsilon \cdot s)) \sum_j \delta_j \epsilon_j s_j}{\pi s_j}\right]^2\right) .$$

Claim 1:

$$\sup_M \left| \frac{\phi_\delta(M, s)}{\pi s_j} \right| \text{ is locally integrable on } \{s \in \Delta_n : \sum_j \delta_j s_j > 0\}$$

(i.e., any point - and thus also any compact subset - of this set has a neighborhood on which the function is integrable).

Proof: Fix one such point  $s_0$ , and consider first a neighborhood  $V_{s_0}$  of  $s_0$  where all strict inequalities among the functions  $\{0; s_1 \dots s_n; (\sum_j \delta_j \epsilon_j s_j)_{\epsilon \in \{-1,1\}^n}\}$  that hold at  $s_0$  are preserved. Let  $\eta > 0$  be strictly smaller than the absolute value at  $s_0$  of any of those functions that does not vanish at  $s_0$ , and assume further that, on

$\tilde{V}_{s_0}$ , all functions that vanish at  $s_0$  remain  $< n$  in absolute value, while all the other functions remain  $> n$  in absolute value. Finally write  $\tilde{V}_{s_0}$  as the finite union of sets  $V_{s_0}$  (and a null set) where on each set  $V_{s_0}$  the ordering of all those functions is constant (and strict). It is sufficient to prove integrability on  $V_{s_0}$ . Assume in particular without loss of generality that, on  $V_{s_0}$ , we have  $0 < s_1 < s_2 < \dots < s_k < n < s_{k+1} < \dots < s_n$  ( $0 < k < n$ ). By assumption  $\exists j > k: \delta_j = 1$ . We have

$$\phi_\delta = \frac{1}{2} \sum_{(\epsilon_1, \dots, \epsilon_n) \in \{-1, 1\}^n} (\ln \frac{M[\delta_1 s_1 + \sum_{j>1} \delta_j \epsilon_j s_j]}{N(s_1, \epsilon_2 s_2, \dots, \epsilon_n s_n)}) - \frac{1}{2} \sum_{(\epsilon_1, \dots, \epsilon_n) \in \{-1, 1\}^n} (\ln \frac{M[-\delta_1 s_1 + \sum_{j>1} \delta_j \epsilon_j s_j]}{N(-s_1, \epsilon_2 s_2, \dots, \epsilon_n s_n)})$$

and we will bound individually of every logarithm in this sum. Let  $f_i(s) = -(-1)^i \delta_1 s_1 + \sum_{j>1} \delta_j \epsilon_j s_j$ ,  $n_i(s) = N(-(-1)^i s_1, \epsilon_2 s_2, \dots, \epsilon_n s_n)$  ( $i = 1, 2$ ). Now  $|\ln((1 + (Mf_1/n_1)^2)/(1 + (Mf_2/n_2)^2))|$  increases monotonically with  $M$  to its limit, so that

$$\sup_M \frac{1}{\pi s_j} \left| \frac{1}{2} \ln \frac{1 + (Mf_1/n_1)^2}{1 + (Mf_2/n_2)^2} \right| = \left| \frac{\ln|f_1/f_2|}{\pi s_j} - \frac{\ln(n_1/n_2)}{\pi s_j} \right|.$$

Thus we only have to show that  $(\ln|f_1/f_2|)/\pi s_j$  and  $(\ln(n_1/n_2))/\pi s_j$  are integrable on  $V_s$ .

For the second term, remark that,  $N$  being a norm, and any two norms on  $\mathbb{R}^n$  being equivalent,  $n_1$  and  $n_2$  are bounded away from 0 and from  $\infty$ , and  $|n_1 - n_2| < N(2s_1, 0, 0, \dots) < Ks_1$ . So  $|\ln n_1/n_2| < K's_1$  - thus we only have to show the integrability of  $s_1/\pi s_j$  on

$$\{s \in \Delta_n : \forall i, s_i > s_1\}.$$

The measure  $\tau$  on  $\Delta_n$  has a bounded density w.r.t.  $ds_1 \dots ds_{n-1}$ , so it is sufficient to prove that  $\int_0^1 \prod_{i=2}^{n-1} \int_{s_1}^1 (ds_i/s_i) < \infty$ , i.e.,  $\int_0^1 |\ln s_1|^{n-2} ds_1 < \infty$ , which is well known.

The term  $(|\ln|f_1/f_2||)/\pi s_j$  appears only if  $\delta_1 = 1$  so assume this. Let  $\Phi(s) = \sum_{j>1} \delta_j \epsilon_j s_j$ : since  $f_1 = \Phi + s_1$ ,  $f_2 = \Phi - s_1$ , we can repeat with  $f_1$  and  $f_2$  the same argument as with  $n_1$  and  $n_2$  if  $\Phi$  does not vanish at  $s_0$ . So assume furthermore  $\Phi(s_0) = \sum_{j>k} \delta_j \epsilon_j s_j^0 = 0$ . Since by assumption  $\sum_{j>k} \delta_j s_j^0 > 0$ , it follows that there exist two indexes  $>k$ , say  $n-1$  and  $n$  (renumbering coordinates), such that  $\delta_{n-1} = \delta_n = 1$ ,  $\epsilon_{n-1} = -1$ ,  $\epsilon_n = 1$ .

Do now the change of coordinates  $(s_1, \dots, s_n) \mapsto ((1/2)f_1, (1/2)f_2, s_2, \dots, s_{n-2})$  using the formulas for  $f_1$  and  $f_2$  and the equation  $\sum s_j = 1$ . Since under our assumptions the change of coordinates has nonzero determinant, it will be sufficient to prove integrability of  $(|\ln|f_1/f_2||)/((f_1 - f_2)^2 s_i)$  on  $1 > s_i > f_1 - f_2 > 0$ ,  $|f_i| < 1/2$ .

Integrating the  $s_i$ 's, this becomes

$$\int \left| \frac{\ln|f_1/f_2|}{f_1 - f_2} \right| \left| \ln|f_1 - f_2| \right|^{n-3} df_1 df_2 \quad \text{on } -\frac{1}{2} < f_2 < f_1 < \frac{1}{2} \text{ or} \\ \text{equivalently on } |f_i| < \frac{1}{2} .$$

Letting  $x_i = |f_i|$ , we get, bounding the integrand,

$$\int \frac{\ln(x_1/x_2)}{x_1 - x_2} \left| \ln|x_1 - x_2| \right|^{n-3} dx_1 dx_2 \quad \text{on } 0 < x_i < \frac{1}{2} ,$$

or, by symmetry around  $x_1 = x_2$ ,

$$\int_{0 < x_1 < x_2 < 1/2} \frac{\ln(x_2/x_1)}{x_2 - x_1} [-\ln(x_2 - x_1)]^k dx_1 dx_2 < \infty, \quad (k > 0)$$

- or, using polar coordinates and increasing slightly the area of integration,

$$\int_{0 < r < 1} \frac{-\ln \tan \theta}{\cos \theta - \sin \theta} [-\ln r - \ln(\cos \theta - \sin \theta)]^k dr d\theta$$

since  $\int_0^1 [-\ln r - A]^k dr$  is a polynomial in  $A$ , we have reduced the problem to showing the finiteness of

$$\int_{0 < \theta < \pi/4} \frac{\ln \cos \theta - \ln \sin \theta}{\cos \theta - \sin \theta} [-\ln(\cos \theta - \sin \theta)]^k d\theta, \quad (k > 0).$$

It is sufficient to show local integrability at  $0$  and  $\pi/4$ ; the ratio being bounded at  $\pi/4$ , it amounts at this point to the well known integrability of  $|\ln x|^n$  near zero; and at  $\theta = 0$ , the argument is just as easy, and reduces to the integrability of  $|\ln x|$  near zero.

This proves the claim.

Q.E.D.

Using now Lebesgue's dominated convergence theorem, we get for any  $n > 0$

$$\lim_{M \rightarrow \infty} \int_{\Delta_n \cap \{s | \sum \delta_i s_i > n\}} \frac{\phi_\delta(M, s)}{\pi s_i} d\tau(s) = \int_{\Delta_n \cap \{s | \sum \delta_i s_i > n\}} \frac{\lim_{M \rightarrow \infty} \phi_\delta(M, s)}{\pi s_i} d\tau(s)$$

and

$$\lim_{M \rightarrow \infty} \phi_\delta(M, s) = \sum_{\epsilon \in \{-1, 1\}^n} \frac{(\pi \epsilon_j) \ln |\sum \delta_j \epsilon_j s_j|}{\Delta_n} - \sum_{\epsilon \in \{-1, 1\}^n} \frac{(\pi \epsilon_j) \ln N(\epsilon \cdot s)}{\Delta_n}$$

if  $\exists j: \delta_j = 0$ , the first sum is zero, so

$$\lim_{M \rightarrow \infty} \phi_\delta(M, s) = I(\delta = (1, 1, \dots, 1)) \sum_{\epsilon \in \{-1, 1\}^n} \frac{(\pi \epsilon_j) \ln |\sum \epsilon_j s_j|}{\Delta_n} - \sum_{\epsilon \in \{-1, 1\}^n} \frac{(\pi \epsilon_j) \ln N(\epsilon \cdot s)}{\Delta_n} .$$

We have also seen in the above proof that both  $(1/\pi s_j) \sum_{\epsilon} (\pi \epsilon_j) \ln |\sum \epsilon_j s_j|$  and  $(1/\pi s_i) \sum_{\epsilon} (\pi \epsilon_j) \ln N(\epsilon \cdot s)$  are integrable over  $\Delta_n$ . (for  $\delta = (1, 1, \dots, 1)$ ,  $\sum \delta_i s_i = 1 > 0$  everywhere); so for  $\delta \neq 0$ ,

$$\begin{aligned} \lim_{M \rightarrow \infty} \int_{\Delta_n} \frac{\phi_\delta(M, s)}{\pi s_j} d\tau(s) &= I(\delta = (1, \dots, 1)) \int_{\Delta_n} \frac{\sum_{\epsilon} (\pi \epsilon_j) \ln |\sum \epsilon_j s_j|}{\pi s_j} d\tau(s) \\ &\quad - \int_{\Delta_n} \frac{\sum_{\epsilon} (\pi \epsilon_j) \ln N(\epsilon \cdot s)}{\pi s_j} d\tau(s) \\ &+ \lim_{n \rightarrow 0} \lim_{M \rightarrow \infty} \int_{\Delta_n \cap \{s | \sum \delta_i s_i < n\}} \frac{\phi_\delta(M, s)}{\pi s_j} d\tau(s) \end{aligned}$$

Therefore, summing over all nonzero  $\delta$ 's,

$$\begin{aligned} P(v_i > 0 \forall i) &= (-1)^{n/2} (2\pi)^{-n} \left[ - \int_{\Delta_n} \frac{\sum_{\epsilon} (\pi \epsilon_j) \ln |\sum \epsilon_j s_j|}{\pi s_j} d\tau(s) \right. \\ &\quad \left. - \int_{\Delta_n} \frac{\sum_{\epsilon} (\pi \epsilon_j) \ln N(\epsilon \cdot s)}{\pi s_j} d\tau(s) \right. \\ &\quad \left. - \sum_{\delta \in \{0, 1\}^n} \sum_{j=1}^n (-1)^j \lim_{n \rightarrow 0} \lim_{M \rightarrow \infty} \int_{\Delta_n \cap \{s | \sum \delta_i s_i < n\}} \frac{\phi_\delta(M, s)}{\pi s_j} d\tau(s) \right] \end{aligned}$$

Let us now compute the last limit.

Assume without loss of generality  $\delta_j = 1$  iff  $j \leq k$ : we want to compute

$$\lim_{n \rightarrow 0} \lim_{M \rightarrow \infty} \int_{\substack{s > 0 \\ \sum s_i = 1 \\ i \leq n \\ \sum s_i \leq n \\ i \leq k}} \frac{\sum (\pi \epsilon_j) \ln [1 + (M(\sum \epsilon_j s_j)/N(\epsilon \cdot s))] }{\prod s_j} d\tau(s)$$

Represent  $s \in \Delta_n$  as  $\alpha x + (1 - \alpha)y$ , with  $x \in \Delta_k$ ,  $y \in \Delta_{n-k}$ ,  $\alpha \in [0,1]$ .

Denote by  $\bar{\tau}_n$  the uniform distribution on  $\Delta_n$ : we have  $\tau = \bar{\tau}_n/(n-1)!$  as noted earlier.

One checks easily that, under  $\bar{\tau}_n$ ,  $\alpha$ ,  $x$  and  $y$  are independent,  $x$  and  $y$  being uniform and  $\alpha$  having the beta-density

$$\frac{(n-1)!}{(k-1)!(n-k-1)!} \alpha^{k-1} (1-\alpha)^{n-k-1} .$$

Thus we get

$$\begin{aligned} & \lim_{n \rightarrow 0} \lim_{M \rightarrow \infty} \int_{\Delta_{n-k}} \frac{d\bar{\tau}_{n-k}(y)}{(n-k-1)!} \int_{\Delta_k} \frac{d\bar{\tau}_k(x)}{(k-1)!} \int_0^\infty \sum (\pi \epsilon_j) \\ & \frac{\ln [1 + (M\alpha(\sum \epsilon_j x_j)/N(\alpha(\epsilon \cdot x) + (1-\alpha)(\epsilon \cdot y)))^2]}{\alpha^k (1-\alpha)^{n-k} \prod x_j \prod y_j} \alpha^{k-1} (1-\alpha)^{n-k-1} d\alpha \\ & = \lim_{n \rightarrow 0} \lim_{M \rightarrow \infty} \int_{\Delta_{n-k}} \frac{d\bar{\tau}_{n-k}(y)}{\prod y_j} \int_{\Delta_k} \frac{d\bar{\tau}_k(x)}{\prod x_j} \int_0^\infty \sum (\pi \epsilon_j) \\ & \ln [1 + (-\frac{M\alpha(\sum \epsilon_j x_j)}{N(\alpha(\epsilon \cdot x) + (1-\alpha)(\epsilon \cdot y))})^2] \frac{d\alpha}{\alpha(1-\alpha)} . \end{aligned}$$

We first want to show that:

Claim 2: The limit (when  $n \rightarrow 0$ ) is not affected if we replace  $N(\alpha(\epsilon \cdot x) + (1 - \alpha)(\epsilon \cdot y))$  by  $N(\epsilon \cdot y)$ , and  $d\alpha/\alpha(1 - \alpha)$  by  $d\alpha/\alpha$ .

A) First replacement

Indeed, as before we get from the equivalence of norms on  $\mathbb{R}^n$  that

$$\left| \ln \frac{N(\alpha(\epsilon \cdot x) + (1 - \alpha)(\epsilon \cdot y))}{N(\epsilon \cdot y)} \right| < K\alpha .$$

Since  $\sup_{A>0} \left| \ln((1 + An_1)/(1 + An_2)) \right| = \left| \ln(n_1/n_2) \right|$ , we get that, after the first replacement, the error in the sum  $\sum_{\epsilon}$  is bounded by  $K \cdot \alpha$  for some  $K > 0$ .

For the same reason, pairing the terms where, for  $j > k$ ,  $\epsilon_j$  is +1 and -1, one finds that, both before and after replacement, the sum  $\sum_{\epsilon}$  is bounded in absolute value by  $K \cdot y_j$ , and for  $j < k$ , one finds the bound

$$Kx_j + K' \left| \ln \left| \frac{\sum_{i \neq j} \epsilon_i x_i + x_j}{\sum_{i \neq j} \epsilon_i x_i - x_j} \right| \right| .$$

Thus, to show that we commit a negligible error in this first replacement when  $n \rightarrow 0$ , we have to show that

$$\int_{\Delta_k} \frac{d\tau_k(x)}{\#x_j} \int_0^{\bar{e}^m} \frac{da}{\alpha} \int_{\Delta_m} \min[\alpha, (x_j + \left| \ln \left| \frac{\epsilon^{j+} \cdot x}{\epsilon^{j-} \cdot x} \right| \right|)^k] \prod_{j=1}^k (y_j)^m \frac{d\tau_m(y)}{\#y_j} < \infty$$

where, for any  $\epsilon \in \{-1, 1\}^k$ ,  $(\epsilon^{j+})_i = \epsilon_i$  for  $i \neq j$ , = +1 for  $i = j$  and  $(\epsilon^{j-})_i = \epsilon_i$  for  $i \neq j$ , = -1 for  $i = j$ .

Let us first bound the inner integral

$$\begin{aligned}
 & \int_{\Delta_m} \min(\beta, (y_j)_{j=1}^m) \frac{d\tau(y)}{\prod y_j} \leq m \int \min(\beta \max_j y_j, (y_j)_{j=1}^m) \frac{d\tau(y)}{\prod y_j} \\
 & \leq m \int_{0 < y_j < 1} \min(\beta \max_j y_j, (y_j)_{j=1}^m) \prod_{j=1}^m \frac{dy_j}{y_j} \\
 & = m^2(m-1) \int_{0 < y_1 < y_m < 1} \min(\beta y_m, y_1) \prod_{j=1}^m \frac{dy_j}{y_j} \\
 & = m^2(m-1) \int_{0 < y_1 < y_m < 1} [\ln(\frac{y_m}{y_1})]^{m-2} \min(y_m^{-1}, \beta y_1^{-1}) dy_1 dy_m \\
 & = m^2(m-1) \int_0^1 dy \int_1^{y^{-1}} [\ln u]^{m-2} \min(\beta, u^{-1}) du .
 \end{aligned}$$

Now the inner integral is

$$\begin{aligned}
 & \text{for } y < \beta \text{ equal to } \beta \int_1^{\beta^{-1}} [\ln u]^{m-2} du + \int_{\beta^{-1}}^{y^{-1}} [\ln u]^{m-2} \frac{du}{u} \\
 & \text{and for } y > \beta \text{ to } \beta \int_1^{y^{-1}} [\ln u]^{m-2} du .
 \end{aligned}$$

Therefore our upper bound equals

$$\begin{aligned}
 & m^2(m-1) \left[ \beta^2 \int_1^{\beta^{-1}} [\ln u]^{m-2} du + \int I(0 < y < u^{-1} < \beta) [\ln u]^{m-2} dy \frac{du}{u} \right. \\
 & \quad \left. + \beta \int I(\beta < y < u^{-1} < 1) [\ln u]^{m-2} du dy \right]
 \end{aligned}$$

$$\begin{aligned}
 &= m^2(m-1) \left[ \beta^2 \int_1^{\beta^{-1}} [\ln u]^{m-2} du + \int_0^{\beta} [\ln v^{-1}]^{m-2} dv + \beta \int_1^{\beta^{-1}} u^{-1} [\ln u]^{m-2} du \right. \\
 &\quad \left. - \beta^2 \int_1^{\beta^{-1}} [\ln u]^{m-2} du \right] \\
 &= m^2(m-1) \left[ \int_0^{\beta} (-\ln v)^{m-2} dv + \frac{\beta}{m-1} (-\ln \beta)^{m-1} \right] \\
 &= m^2 \int_0^{\beta} (-\ln v)^{m-1} dv
 \end{aligned}$$

by intergration by parts. The same integration by parts gives by induction that this last integral equals

$$(m-1)! \beta \sum_{i=0}^{m-1} \frac{(-\ln \beta)^i}{i!} .$$

Therefore, since we are only interested in values of  $\beta < e^{-m} < 1$ , we get

$$\int_{\Delta_m} \text{Min}(\beta, (y_j)_{j=1}^m) \frac{d\tau(y)}{\prod y_j} < K_m \beta |\ln \beta|^{m-1} .$$

Let  $F_m(\beta) = \beta |\ln \beta|^m$ : we thus have to prove that

$$\int_{\Delta_k} \frac{d\tau(x)}{\prod x_j} \int_0^{-m} F_{m-1} \left( \text{Min}[\alpha, (x_j + \left| \ln \left| \frac{\epsilon^{j+} \cdot x}{\epsilon^{j-} \cdot x} \right| \right|)_{j=1}^k] \right) \frac{d\alpha}{\alpha} < \infty .$$

To evaluate the inner integral, let for short  $y_j = x_j + \left| \ln \left| (\epsilon^{j+} \cdot x) / (\epsilon^{j-} \cdot x) \right| \right|$  then the inner integral is bounded by

$$\min_j \int_0^{e^{-m}} F_{m-1}(\min(\alpha, y_j)) \frac{d\alpha}{\alpha}$$

$F_{m-1}$  being increasing in  $[0, e^{-m}]$ . Call this last integral  $\phi(y)$ ; we get letting  $\rho = \min(y, e^{-m})$

$$\begin{aligned}\phi(y) &= \int_0^{\rho} (|\ln x|^{m-1}) dx - F_{m-1}(\rho)(\ln \rho + m) \\ &\leq K_m F_{m-1}(\rho) - (\ln \rho + m)F_{m-1}(\rho) \\ &\leq K'_m F_m(\rho)\end{aligned}$$

(using our previous bound for the integral) .

Thus it will be sufficient to show that, letting  $\psi(y) = F_m(\min(y, e^{-m}))$

$$\int_{\Delta_k \cap \{x_1 < x_i\}} \psi(y_1) \frac{d\tau(x)}{\prod x_i} < \infty .$$

Since  $u > 0, v > 0$  implies  $\psi(u+v) < \psi(u) + \psi(v)$ , it will be sufficient to show separately that

$$\int_{x_1 < x_i} |\ln x_1|^m \frac{d\tau(x)}{\prod x_i} < \infty$$

and that

$$\int_{x_1 < x_i} \psi\left(\left|\ln \frac{\epsilon^{1+}}{\epsilon^{1-}} \cdot x\right|\right) \frac{d\tau(x)}{\prod x_i} < \infty .$$

The first integral is bounded by - letting  $\|x\|_\infty = \max|x_i|$ , and  $\tau'$  being Lebesgue measure on  $\{x | \|x\|_\infty = 1\}$  -

$$\begin{aligned}
 & \int_{x_1 < x_i} \left[ \ln \frac{k \cdot \|x\|_\infty}{x_1} \right]^m \frac{d\tau(x)}{\prod x_i} < (k-1) \int_{x_1 < x_i < x_2} \left[ \ln \frac{kx_2}{x_1} \right]^m \frac{d\tau'(x)}{\prod x_i} \\
 &= C_k \int_{0 < x_1 < x_i < x_2 = 1} \left[ \ln \frac{kx_2}{x_1} \right]^m dx_1 \prod_{i>2} \frac{dx_i}{x_i} \\
 &= C_k \int_0^1 \left( \ln \frac{k}{x_1} \right)^m \left( \ln \frac{1}{x_1} \right)^{k-2} dx_1 < \infty .
 \end{aligned}$$

For the second integral, we will prove local integrability, i.e., that, for any  $x \in \Delta_k \cap \{x_1 < x_i\}$  there is a neighborhood of  $x$  in this set where the function is integrable.

If  $x_1 > 0$ , then  $x_i > 0 \forall i$  so that the integrand is locally bounded. Otherwise, one has  $\epsilon \cdot x = \epsilon^{1+} \cdot x = \epsilon^{1-} \cdot x$ : if  $\epsilon \cdot x \neq 0$ , then locally  $|\ln(\epsilon^{1+} \cdot x)/(\epsilon^{1-} \cdot x)| \leq Kx_1$ , so  $\psi \leq K' \psi(x_1)$ , and we have just shown this bound to be integrable.

Thus there just remains to consider the case where  $\epsilon^{1+} \cdot x = \epsilon^{1-} \cdot x = 0$ .

Since  $x_1 = 0$ ,  $\sum x_i = 1$ , there exists an index  $j \neq 1$  with  $x_j > 1/k$ , and since further  $\epsilon \cdot x = 0$ , there exists another index  $j' \neq 1$  with  $x_{j'} > k^{-2}$ , and  $\epsilon_j \epsilon_{j'} = -1$ . Assume without loss of generality that  $j' = k-1$ ,  $j = k$ , and make the change of variables

$$(x_1 \dots x_k) + (f_1, f_2, x_2 \dots x_{k-2})$$

using the equations

$$\epsilon^{1+} \cdot x = f_1 , \quad \epsilon^{1-} \cdot x = f_2 , \quad \sum x_i = 1 .$$

The integrability on  $(\psi/\prod x_i) d\tau(x)$  is equivalent to that of  $\psi/(\prod_{i<k-1} x_i)$ , which is equivalent to that of

$$\frac{\psi(|\ln|f_1/f_2||)}{(f_1 - f_2)} df_1 df_2 \prod_{i=2}^{k-2} \left( \frac{dx_i}{x_i} \right)$$

over  $\{|f_i| < 1/2, 0 < f_1 - f_2 < x_i < 1\}$  - or, integrating over the  $x_i$ 's:

$$\frac{|\ln(f_1 - f_2)|^\ell \psi(|\ln|f_1/f_2||)}{f_1 - f_2} df_1 df_2 .$$

The integrand is only increased if we replace  $f_1 - f_2$  by  $|f_1| - |f_2|$  - so we assume  $0 < f_i < 1$ , inserting absolute value of differences.

Further by symmetry it is sufficient to consider the case

$f_1 > f_2$ :

$$\int_{0 < f_2 < f_1 < 1} \frac{[-\ln(f_1 - f_2)]^\ell F_m[\ln((f_1/f_2) \wedge (1 + \delta_m))]}{f_1 - f_2} df_1 df_2 .$$

Let  $f_1 = y, f_2/f_1 = 1 - x$ : our integral becomes

$$\int_0^1 \frac{F_m[\ln((1-x)^{-1} \wedge (1 + \delta))]^\ell}{x} dx \int_0^1 [-\ln(xy)]^\ell dy ;$$

the inner integral is  $(1/x) \int_0^x [-\ln z]^\ell dz$ , which by a previous computation is equal to a polynomial in  $[-\ln x]$ : everything amounts to showing that

$$\frac{[-\ln x]^{\frac{k}{m}} F_m[\ln((1-x)^{-1} \wedge (1+\delta_m))]}{x} dx$$

is integrable on  $[0,1]$ . The integrand is bounded except at  $x = 0$ , where it is bounded by  $([-\ln x]^{\frac{k}{m}} F_m(2x))/x$ , i.e., a polynomial in  $(-\ln x)$  - this we know to be integrable.

Thus, we finished proving that the first replacement can be made.

### B) Second Replacement

Once this first replacement is done, the sum in the integrand is, by our previous argument, bounded by

$$\frac{K}{\epsilon} \min \left\{ (y_j)^{n-k}, \left( \left| \ln \left| \frac{\epsilon^{1+} \cdot x}{\epsilon^{1-} \cdot x} \right| \right|^k \right)_{j=1}^k \right\} .$$

So, to show that the second replacement can be done, we have to show that this function is integrable for

$$\frac{d\tau(y)}{dy_j} \cdot \frac{d\tau(x)}{dx_j} \cdot \frac{1}{1-\alpha} \quad (\text{since } \alpha(1-\alpha) = \frac{1}{\alpha} + \frac{1}{1-\alpha})$$

over say  $\alpha < 1/2$  - thus that

$$\int_0^{\frac{1}{2}} \frac{d\tau(x)}{dx_j} d\alpha / \frac{d\tau(y)}{dy_j} \min \left\{ (y_j)^{n-k}, \left( \left| \ln \left| \frac{\epsilon^{1+} \cdot x}{\epsilon^{1-} \cdot x} \right| \right|^k \right)_{j=1}^k \right\} < \infty .$$

For the first integral we can use our previous computation, and the integral over  $\alpha$  disappears, so we are left to prove that

$$\int_{x_1 \leq x_j} \frac{F_m \left( \left| \ln \left| \frac{\epsilon^{1+} \cdot x}{\epsilon^{1-} \cdot x} \right| \right| \wedge e^{-m} \right)}{\frac{1}{dx_j}} d\tau(x) < \infty$$

and this we have shown previously also. So Claim 2 is proved. Q.E.D.

Thus we have to compute

$$\lim_{n \rightarrow 0} \lim_{M \rightarrow \infty} \frac{1}{2} \int \frac{d\tau_{n-k}(y)}{\pi y_j} \int \frac{d\tau_k(x)}{\pi x_j} \sum_{0}^M \sum_{\epsilon^x, \epsilon^y} (\Pi \epsilon_j^x)(\Pi \epsilon_j^y) \ln [1 + [Ma(\epsilon^x \cdot x)/N(0 \cdot x, \epsilon^y \cdot y)]^2] \frac{da}{a}$$

Letting  $Ma = u$ , this becomes

$$\lim_{n \rightarrow 0} \lim_{M \rightarrow \infty} \frac{1}{2} \int \frac{d\tau_{n-k}(y)}{\pi y_j} \int \frac{d\tau_k(x)}{\pi x_j} \sum_{0}^M \sum_{\epsilon^x, \epsilon^y} (\Pi \epsilon_j^x)(\Pi \epsilon_j^y) \ln [1 + [u(\epsilon^x \cdot x)/N(0 \cdot x, \epsilon^y \cdot y)]^2] \frac{du}{u} .$$

Now, the limit when  $M$  goes to infinity becomes independent of  $n$ , so we get, using for short  $\epsilon$  for  $\epsilon^x$  and  $n$  for  $\epsilon^y$ :

$$\lim_{M \rightarrow \infty} \frac{1}{2} \int \frac{d\tau_{n-k}(y)}{\pi y_j} \int \frac{d\tau_k(x)}{\pi x_j} \sum_{0}^M \sum_{\epsilon, n} (\Pi \epsilon_j)(\Pi n_j) \ln [1 + [u(\epsilon \cdot x)/N(0 \cdot x, n \cdot y)]^2] \frac{du}{u} .$$

Denote the inner integral by  $\phi_M(x, y)$ . We have

$$\begin{aligned} \phi_M(x, y) &= \sum_{\epsilon, n} (\Pi \epsilon_j)(\Pi n_j) \int_0^{M|\epsilon \cdot x|/N(0 \cdot x, n \cdot y)} \ln(1 + v^2) \frac{dv}{v} \\ &= \sum_{\epsilon, n} (\Pi \epsilon_j)(\Pi n_j) \int_0^{|\epsilon \cdot x|/N(0 \cdot x, n \cdot y)} \ln(1 + M^2 w^2) \frac{dw}{w} \\ &= \int_0^\infty \sum_{\epsilon, n} (\Pi \epsilon_j)(\Pi n_j) I[w < |\epsilon \cdot x|/N(0 \cdot x, n \cdot y)] \ln(1 + M^2 w^2) \frac{dw}{w} . \end{aligned}$$

If instead of  $\ln(1 + M^2 w^2)$  in this integral we had a constant, say  $\ln M^2$ , the integral would still exist because the integrand vanishes in a neighborhood of the origin, and would have a value equal to  $\ln M^2$  times

$$\sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) \ln |\epsilon + x| - \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) \ln N(0 + x, n + y) = 0$$

- the first term being zero because of  $\sum_n (\Pi n_j)$  and the second because  $\sum_{\epsilon} (\Pi \epsilon_j)$ .

So we still have

$$\phi_M(x, y) = \int_0^\infty \left[ \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) I[w < |\epsilon + x| / N(0 + x, n + y)] \right] \ln(w^2 + M^{-2}) \frac{dw}{w} .$$

Now the integrand is uniformly bounded, and vanishes outside some closed interval disjoint from zero, so that by Lebesgue's dominated convergence theorem

$$\begin{aligned} \lim_{M \rightarrow \infty} \phi_M(x, y) &= \int_0^\infty \left[ \sum_{\epsilon, n} \right] (\ln w^2) \frac{dw}{w} = \int_{w=0}^\infty \left[ \sum_{\epsilon, n} \right] 2 \ln w d\ln w \\ &= \int_{w=0}^\infty \left[ \sum_{\epsilon, n} \right] d\ln^2 w = \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) \ln^2 \frac{|\epsilon + x|}{N(0 + x, n + y)} \\ &= \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) \ln^2 |\epsilon + x| + \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) \ln^2 N(0 + x, n + y) \\ &\quad - 2 \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) \ln |\epsilon + x| \ln N(0 + x, n + y) \\ &= -2 \left[ \sum_{\epsilon} (\Pi \epsilon_j) \ln |\epsilon + x| \right] \left[ \sum_n (\Pi n_j) \ln N(0 + x, n + y) \right] \end{aligned}$$

(the first of the three sums is zero because of  $\sum_n (\Pi n_j)$  and the second because  $\sum_\epsilon (\Pi \epsilon_j)$ ).

Now we have to show that when we apply the first two integrations to  $\phi_M - \lim_M \phi_M$ , we get something going to zero, i.e., that

Claim 3:

$$\lim_{M \rightarrow \infty} \int \frac{d\tau_{n-k}(y)}{\Pi y_j} \int \frac{d\tau_k(x)}{\Pi x_j} \int_0^\infty \left| \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) I[w < |\epsilon + x| / N(0 + x, n + y)] \right| \ln(1 + (Mw)^{-2}) \frac{dw}{w} = 0 .$$

By Lebesgue's dominated convergence theorem, since  $\ln(1 + (Mw)^{-2})$  decreases pointwise to zero, it will be sufficient to show that

$$\int \frac{d\tau_{n-k}(y)}{\Pi y_j} \int \frac{d\tau_k(x)}{\Pi x_j} \int_0^\infty \left| \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) I[w < |\epsilon + x| / N(0 + x, n + y)] \right| \ln(1 + w^{-2}) \frac{dw}{w} < \infty$$

or, replacing  $w$  by  $z^{-1/2}$ , that

$$\int \frac{d\tau_{n-k}(y)}{\Pi y_j} \int \frac{d\tau_k(x)}{\Pi x_j} \int_0^\infty \left| \sum_{\epsilon, n} (\Pi \epsilon_j) (\Pi n_j) I[z|\epsilon + x|^2 > N^2(0 + x, n + y)] \right| \ln(1 + z) \frac{dz}{z} < \infty .$$

If  $z$  is close enough to zero (smaller than  $\min_y N^2(0 + x, y)$ ) then the sum  $\sum_{\epsilon, n}$  is identically zero (i.e., for all  $x$  and  $y$ ), so we can replace  $dz/z$  by  $dz/(1 + z)$ , and get

$$\int \frac{d\tau_{n-k}(y)}{\pi y_j} \int \frac{d\tau_k(x)}{\pi x_j} \sum_{\epsilon, n}^{\infty} \left| \sum_{\epsilon, n} (\pi \epsilon_j) (\pi n_j) I[z|\epsilon + x|^2 > N^2(0 + x, n + y)] \right|$$

$$d \ln^2(1 + z) < \infty ?$$

Let us now try to bound the sum  $\sum_{\epsilon, n}$ .

Pairing the terms where  $n_1$  has opposite signs we get

$$\sum_{\epsilon, n} = \sum_{\epsilon, n} (\pi \epsilon_j) (\pi n_j) I(n_+ < z | \epsilon, x |^2 < n_-)$$

using  $n_{\pm}$  for  $N^2(0 + x, \pm n_1 y_1, n_2 y_2, n_3 y_3, \dots)$ .

Pairing now the terms where  $\epsilon_1$  has opposite signs, we get,

$$\text{letting } u = \sum_{j>1} \epsilon_j x_j, u_+ = |u + \epsilon_1 x_1|^2, u_- = |u - \epsilon_1 x_1|^2:$$

$$\begin{aligned} \sum_{\epsilon, n} = & \sum_{\epsilon, n} (\pi \epsilon_j) (\pi n_j) \{ [I(\frac{n_-}{n_+} < \frac{u_+}{u_-}) + I(\frac{n_-}{n_+} < \frac{u_-}{u_+})] I(\frac{u_+}{u_-} < z < \frac{n_-}{n_+}) \\ & + I(\frac{u_+}{u_-} < \frac{n_-}{n_+}) I(\frac{n_-}{n_+} < z < \frac{n_+}{u_-}) + I(\frac{u_-}{u_+} < \frac{n_-}{n_+}) I(\frac{n_-}{u_+} < z < \frac{n_-}{n_+}) \} . \end{aligned}$$

Thus we get by integrating

$$\begin{aligned} \int \sum_{\epsilon, n} d \ln^2(1 + z) & \leq \sum_{\epsilon, n} \{ [I(\frac{n_-}{n_+} < \frac{u_+}{u_-}) + I(\frac{n_-}{n_+} < \frac{u_-}{u_+})] \Delta_1 + I(\frac{u_+}{u_-} < \frac{n_-}{n_+}) \Delta_2 \\ & + I(\frac{u_-}{u_+} < \frac{n_-}{n_+}) \Delta_3 \} \end{aligned}$$

where

$$\begin{aligned} \Delta_1 &= I(n_- > n_+) (\ln^2(1 + \frac{n_-}{u_+}) - \ln^2(1 + \frac{n_+}{u_+})) \leq I(n_- > n_+) (\ln \frac{n_- + u_+}{n_+ + u_+}) \\ & (\ln(1 + n_-) - \ln u_+) + 2 \leq K I(n_- > n_+) (\ln \frac{n_-}{n_+}) (1 - \ln u_+) \end{aligned}$$

$$\Delta_2 = I(u_+ > u_-) \left( \ln^2 \left( 1 + \frac{n_+}{u_-} \right) - \ln^2 \left( 1 + \frac{n_+}{u_+} \right) \right) < I(u_+ > u_-)$$

$$\left( \ln \frac{1/n_+ + 1/u_-}{1/n_+ + 1/u_+} \right) \left( \ln \left( 1 + n_+ \right) - \ln u_- \right) \cdot 2 < K I(u_+ > u_-) \left( \ln \frac{u_+}{u_-} \right) \left( 1 - \ln u_- \right)$$

and similarly

$$\Delta_3 < K I(u_+ < u_-) \left( \ln \frac{u_-}{u_+} \right) \left( 1 - \ln u_+ \right) .$$

Remark that  $\Delta_2$  is used only when  $1 < u_+/u_- < n_-/n_+ < K_1$ , so that, by modifying the constant  $K$ , one can replace the factor  $(1 - \ln u_-)$  by  $(1 - \ln u_+)$ .

Obviously this formula remains valid - readjusting  $K$  - when re-interpreting  $n_{\pm}$  (resp.  $u_{\pm}$ ) as previous  $n_{\pm}$  (resp.  $u_{\pm}$ ) .

Remark also that in all three cases, one uses the smaller of the factors  $|\ln(u_+/u_-)|$ ,  $|\ln(n_-/n_+)|$ . Thus we get simply

$$\int \left| \sum_{\epsilon, n} \left| d \ln^2(1+z) \right| \right| < K \sum_{\epsilon, n} I(n_- > n_+) [\min(\ln \frac{n_-}{n_+}, |\ln \frac{u_+}{u_-}|)] (1 - \ln u_+)$$

As already remarked before, the equivalence of norms in  $\mathbb{R}^n$  implies that  $\ln(n_-/n_+) < K'y_1 (\ll K')$ . In particular, if we assume for instance  $u_+ > u_-$ , we can replace  $|\ln(u_+/u_-)|$  by  $\ln[\min((u_+/u_-), e^{K'})] < K''((u_+ - u_-)/u_+)$ . Now  $u_+ > u_-$  is equivalent to  $\epsilon_1 u > 0$ , so  $u_+ - u_- = \epsilon_1 u + x_1 - |\epsilon_1 u - x_1| = 2 \min(\epsilon_1 u, x_1) = 2 \min(|u|, x_1)$ , and  $u_+ = |u| + x_1$ . So we can replace  $|\ln(u_+/u_-)|$  by  $(\min(|u|, x_1)) / (|u| + x_1)$  if  $u_+ > u_-$ , and thus also in the dual case. Thus we get

$$\int \left| \sum_{\epsilon, n} \left| d \ln^2(1+z) < c \sum_{\epsilon} (y_1 \wedge \frac{x_1 \wedge |u|}{|u| + x_1}) (1 - \ln|\langle \epsilon, x \rangle|) \right. \right| .$$

But  $y_1$  could have been any  $y_j$ , and in particular their minimum  $\wedge y_j$ . Using now our previous formula

$$\int_{\Delta_m} \text{Min}(\beta, (y_j)_{j=1}^m) \frac{d\tau(y)}{\prod y_j} < K_m \beta |\ln \beta|^{m-1}$$

we get

$$\int_{\Delta_m} \frac{d\tau(y)}{\prod y_j} \int \left| \sum_{\epsilon, n} \left| d \ln^2(1+z) < c' \sum_{\epsilon} (1 - \ln|\langle \epsilon, x \rangle|) y |\ln y|^{m-1} \right. \right| .$$

where

$$y = \frac{x_1 \wedge |u|}{x_1 + |u|} .$$

We have to show that this is integrable  $d\tau(x)/\prod x_j$ , at all points of  $\Delta_k \cap \{x_1 < x_i \forall i\}$  - if another coordinate was minimal, let this play the role of  $x_1$ .

There is no problem if  $x_1 > 0$  because  $\ln|\langle \epsilon, x \rangle|$  is integrable for Lebesgue measure, and  $y |\ln y|^{m-1}$  is bounded.

Fix now an  $\epsilon$ . If  $x_1 = 0$ , and  $|u| > 0$ , then the factor  $(1 - \ln|\langle \epsilon, x \rangle|)$  is locally bounded, and  $y$  is locally of the order of  $x_1$ , so that we have to show the integrability of  $x_1 |\ln x_1|^r (d\tau(x)) / \prod x_i$  on  $\{x_1 < x_i\}$  which we have already done before.

There remains thus only the case where  $x_1 = u = 0$ . In that case, as we argued already before, there exists two different coordinates  $j$  and  $j'$ , different from 1, such that  $x_j > 0, x_{j'} > 0, \epsilon_j \epsilon_{j'} = -1$ . We

can assume without loss of generality that  $j = k$ ,  $j' = k - 1$ , and can change coordinates

$$x_1 \dots x_k \rightarrow u, x_1 \dots x_{k-2}$$

using the equations  $\sum_{i=2}^k \epsilon_i x_i = u$ ,  $\prod_{i=1}^k x_i = 1$ .

We therefore have in effect to prove that - assuming without loss of generality that  $\epsilon_1 = 1$ :

$$\int_{\substack{|u| \leq 1 \\ 0 \leq x_1 \leq x_i \leq 1}} (1 - \ln|u + x_1|) \frac{x_1 \wedge |u|}{x_1 + |u|} \ln^r(\frac{x_1 + |u|}{x_1 \wedge |u|}) du \prod_{i=1}^{k-2} \frac{dx_i}{x_i} < \infty$$

or, integrating over  $x_i$  for  $i > 1$ :

$$\int_{\substack{|u| \leq 1 \\ 0 \leq x \leq 1}} (1 - \ln|u + x|) \frac{x \wedge |u|}{x + |u|} \ln^r(\frac{x + |u|}{x \wedge |u|}) |\ln x|^{k-3} du \frac{dx}{x} < \infty$$

or

$$\int_0^1 \int_0^1 [1 - \ln|x - u|] \frac{x \wedge u}{x + u} \ln^r(\frac{x + u}{x \wedge u}) |\ln x|^s du \frac{dx}{x} < \infty$$

Replacing  $|\ln x|^s$  by  $|\ln(x \wedge u)|^s$ , and  $dx/x$  by  $dx/(x \wedge u)$ , one sees it is sufficient to consider  $x < u$ :

$$\int I(0 < x < u < 1) [1 - \ln(u - x)] \frac{1}{x + u} \ln^r(1 + \frac{u}{x}) |\ln x|^s du dx < \infty ?$$

Since  $\ln^r(1 + u/x)$  can be written as a polynomial in  $\ln(x + u)$  and  $\ln u$ , and since  $|\ln(x + u)| < |\ln(u - x)|$ , the whole thing amounts to proving that

$$\int_{0 < x < u < 1} |\ln(u - x)|^r |\ln x|^s \frac{du dx}{u + x} < \infty \quad (\text{whatever be } r > 0, s > 0)$$

or, letting  $z = x/u$

$$\int_0^1 \frac{dz}{1+z} \int_0^1 (-\ln u - \ln(1-z))^r (-\ln u - \ln z)^s du < \infty$$

the integrand in the second integral is a polynomial in  $\ln u$ , whose coefficients are polynomials in  $\ln z$  and  $\ln(1-z)$ .

Since any power of  $\ln u$  is integrable, the first integral yields a polynomial in  $\ln z$  and  $\ln(1-z)$ ; since  $1/(1+z)$  is bounded, the outer integral boils down to

$$\left(\int_0^{1/2} + \int_{1/2}^1\right) |\ln z|^r |\ln(1-z)|^s dz ,$$

which is finite for the same reason.

This finishes the proof of Claim 3.

Q.E.D.

It follows that

$$\begin{aligned} \lim_{M \rightarrow \infty} \int \frac{d\tau_{n-k}(y)}{\Pi y_j} \int \frac{d\tau_k(x)}{\Pi x_j} \phi_M(x, y) &= \int \frac{d\tau_{n-k}(y)}{\Pi y_j} \int \frac{d\tau_k(x)}{\Pi x_j} \left( \lim_{M \rightarrow \infty} \phi_M(x, y) \right) \\ &= -2 \left[ \int_{\Delta_k} \frac{\sum (\Pi \epsilon_j) \ln |\langle \epsilon, x \rangle|}{\Pi x_j} d\tau_k(x) \right] \left[ \int_{\Delta_{n-k}} \frac{\sum (\Pi n_j) \ln N(0, x, n, y)}{\Pi y_j} d\tau_{n-k}(y) \right] \end{aligned}$$

and therefore that

$$\lim_{n \rightarrow 0} \lim_{M \rightarrow \infty} \int_{\substack{s \in \Delta_n, \\ i < k}} \frac{1}{2} \frac{\sum (\pi \epsilon_j) \ln [1 + (M(\sum_{j < k} \epsilon_j s_j)/N(\epsilon \cdot s))^2]}{\pi s_j} d\tau_n(s)$$

$$= - A_k \int_{\Delta_{n-k}} \frac{\sum (\pi n_j) \ln N(0 \cdot x, n \cdot y)}{\pi y_j} d\tau_{n-k}(y)$$

where

$$A_k = \int_{\Delta_k} \frac{\sum (\pi \epsilon_j) \ln |\langle \epsilon, x \rangle|}{\pi x_j} d\tau_k(x).$$

If we use also  $A_0 = -1$ , we get therefore

$$P(v_i > 0 \forall i) = (-1)^{n/2} (2\pi)^{-n} [-A_n + A_0 \int_{\Delta_n} \frac{\sum (\pi n_j) \ln N(n \cdot y)}{\pi y_j} d\tau_n(y) + \sum_{\substack{\delta \in \{0,1\}^n, n_i \delta_i > 0 \\ \delta \neq (0,\dots,0), (1,\dots,1)}} A_{\sum \delta_j} \int_{\substack{y \in \Delta_n \\ \langle \delta, y \rangle = 0}} \frac{\sum_{j: \delta_j = 1} (\pi n_j) \ln N(n \cdot y)}{\pi y_j} d\tau_{n-\sum \delta_j}(y)]$$

(remarking that  $A_k = 0$  if  $k$  is odd). Thus:

$$P(v_i > 0 \forall i) = (-1)^{n/2} (2\pi)^{-n} \sum_{\delta \in \{0,1\}^n} A_{\sum \delta_j} \int_{\substack{|y|=1 \\ y_j: \delta_j y_j = 0}} \frac{\ln N(y)}{\pi y_j} d\tau_{n-\sum \delta_j}(y)$$

where

•  $\tau_{n-\sum \delta_j}$  is Lebesgue measure on the corresponding set,

• the integrals are Cauchy principal values

. for  $\delta = (1, \dots, 1)$ , the integral over the empty set is set to -1

$$\cdot A_0 = -1, A_k = \int_{\Delta_k} \frac{\sum_{\epsilon \in \{-1, 1\}^k} (\prod \epsilon_j)^{1/n} |\langle \epsilon, x \rangle|}{\prod x_j} d\tau_k(x) .$$

If each  $v_i$  has norm 2, total mass zero, and all  $v_i$  are mutually singular,  $P(v_i > 0 \forall i) = 2^{-n}$  obviously. Also  $R_v = [-1, 1]^n$ , so that  $N(y) = \sup_{x \in [-1, 1]^n} \langle x, y \rangle = \sum |y_i| = \|y\|$ .

Since we integrate on the unit ball, it follows that in this case our equation yields

$$2^{-n} = (-1)^{n/2} (2\pi)^{-n} [ -A_n ]$$

thus

$$A_n = -(-1)^{n/2} \pi^n$$

thus

$$\begin{aligned} (-1)^{n/2} (2\pi)^{-n} A_k &= -2^{-n} [ (-1)^{n/2} \pi^{-n} (-1)^{-k/2} \pi^k ] \\ &= -2^{-n} [ (-1)^{-(n-k)/2} \pi^{-(n-k)} ] \\ &= \frac{2^{-n}}{A_{n-k}} . \end{aligned}$$

Thus:

$$P(v_i > 0 \forall i) = 2^{-n} [ 1 - \sum_{\substack{\delta \in \{0, 1\}^n \\ \delta \neq 0 \\ \sum \delta_j \text{ even}}} \frac{(-1)^{\sum \delta_j / 2}}{\pi^{\sum \delta_j}} \int_{\substack{\|y\|=1 \\ \forall j : (1-\delta_j)y_j = 0}} \frac{\ln N(y)}{\prod y_j} d\tau_{\sum \delta_j}(y) ]$$

This formula is valid for the case  $n$  even, say  $n = 2k$ . But  $P(v_1 \dots v_{2k-1} > 0) = P(v_1 \dots v_{2k} > 0) + P(v_1, v_2, \dots, v_{2k-1}, -v_{2k} > 0)$  can be computed from this formula, and yields then the same formula with  $n = 2k - 1$ : thus the formula is valid for all  $n \geq 1$ .

Thus, for all  $\#I \geq 1$ :

$$P(v_i > 0 \forall i \in I) = 2^{-(\#I)} [1 - \sum_{\substack{\emptyset \neq J \subset I \\ [\#J \text{ even}]}} \frac{1}{(\pi_i)^{\#J}} \int_{\substack{y \in R \\ \|y\|=1}} \frac{\ln N_J(y)}{\prod_{j \in J} y_j} d\tau(y)]$$

where

$$N_J(y) = \sup_{\|x\|<1} \sum_{j \in J} y_j v_j(x).$$

Remark that, by the symmetry of the norm ( $N_J(y) = N_J(-y)$ ), the restriction to  $\#J$  even is not necessary: the integral will be zero for  $\#J$  odd.

The integrals have to be understood as Cauchy principal values, in the following sense: define a set  $C \subseteq R^J$  to be symmetric iff  $y \in C \Leftrightarrow (|y_j|)_{j \in J} \in C$ ; say that  $C$  consists only of non zero elements iff  $y \in C \Rightarrow y_j \neq 0 \forall j \in J$ . Then the integral is to be understood as the limit of the integrals over an arbitrary sequence of closed symmetric sets  $C_i$  consisting only of non zero elements, and such that the measure of the complement of  $C_i$  goes to zero.

The norm  $\|y\|$  used to derive the formula was the  $\ell_1$ -norm  $\sum |y_j|$ , but the formula of change of variables for surface measures yields now that it remains valid for any norm  $\|\cdot\|$  on  $R^J$  such that  $\|(y_1 \dots y_k)\| = \|( |y_1|, |y_2|, \dots, |y_k| )\|$ .

The same formula permits us to rewrite our expression using the surface measure  $\sigma_J$  on the unit sphere of the norm  $N_J$ :

$$P(v_i(x) > 0 \forall i \in I) = 2^{-\#I} [1 + \sum_{\substack{\emptyset \neq J \subset I \\ [\#J \text{ even}]}} \frac{1}{(\pi i)^{\#J}} \int \frac{\ln \|y\|}{\prod_j y_j} d\sigma_J(y)] .$$

Remark: Obviously the formula we got is not very transparent - this may be due to the fact that it has to reflect the peculiar geometry of the positive orthant. It would therefore be interesting to have also an expression for the density over directions - i.e., on projective space.

#### SECTION 4: To Mess Everything Up: Some Extension Possibilities

##### 4.1 Extension of the Cylinder Measure

Given a cylinder measure  $\mu$  on a locally convex space  $E$  with dual  $E'$ , one can use Kolmogorov's existence theorem for a projective limit of measures as done in the proof of Theorem 1, and a Hamel basis of  $E'$ , to obtain an equivalent characterization of  $\mu$  as a countably additive measure on the Baire  $\sigma$ -field of the weak completion  $\bar{E}$  of  $E$ , using also a recent result of Edgar.<sup>3/</sup>

Using this, one can then best define the corresponding integral in the following way: let  $\alpha$  vary in the increasing net of all finite subsets of  $E'$ . For any  $\alpha$ , and any  $x \in E \setminus E$ , let  $V_\alpha(x) = \{y \in E' : y = \phi(x) \forall \phi \in \alpha\}$ . For any bounded function  $f$  on  $E$ , define its extension  $\tilde{f}$  to  $\bar{E}$  by  $\tilde{f}(x) = \lim_{\alpha} \sup_{y \in V_\alpha(x)} f(y)$  at all  $x \in E \setminus E$ .

Finally define the upper integral  $\bar{\mu}(f)$  as the upper integral of  $\bar{f}$  for the countably additive measure  $\mu$  on the Baire  $\sigma$ -field of  $E$ .

Given  $\bar{\mu}$ , one can use finitely additive integration theory in the standard way (cfr. for instance Dunford and Schwartz, Linear Operators, Part I). More precisely, one has:

- 1)  $\bar{\mu}(1) = 1, \bar{\mu}(-1) = -1, \bar{\mu}$  is monotonic;
- 2)  $a > 0$  implies  $\bar{\mu}(af) = a\bar{\mu}(f)$ ;
- 3)  $\bar{\mu}(f + g) \leq \bar{\mu}(f + g) + \bar{\mu}(f \wedge g) \leq \bar{\mu}(f) + \bar{\mu}(g)$  whenever  $\bar{\mu}(f) < +\infty, \bar{\mu}(g) < +\infty$  (the first inequality is subadditivity, the second follows from the corresponding formula for upper integrals,

and from  $\bar{f} \vee \bar{g} = \overline{f \vee g}, \bar{f} \wedge \bar{g} = \overline{f \wedge g}$ ;

- 4)  $\bar{\mu}(f \vee (-n)) \rightarrow \bar{\mu}(f) \neq f, \bar{\mu}(f \wedge 0) > -\infty \Rightarrow \bar{\mu}(f) = \lim_{n \rightarrow \infty} \bar{\mu}(f \wedge n)$ .

Those properties immediately imply that  $L = \{f; \bar{\mu}(f) + \bar{\mu}(-f) < 0\}$  is a vector lattice containing the constants, and that  $\mu$  is a positive linear functional on  $L$ . Hence  $A = \{A | I_A \in L\}$  is a Boolean algebra and  $\bar{\mu}$  a finitely additive probability on  $A$ . Hence  $f \in L$  and  $s < t$  imply  $\mu_*\{f > s\} \geq \mu^*\{f > t\}$  (reduce to  $s = 0 < f < t = 1$ , then  $\int f d\mu$  is in between - we use  $\mu_*(A) = \sup\{\bar{\mu}(f) | f \in L, f \leq I_A\}$ , and  $\mu^*(A) = \inf\{\bar{\mu}(f) | f \in L, f \geq I_A\}$ ). Therefore, if  $f \in L$ , then for all but countably many  $t$ 's,  $\mu^*\{f > t\} = \mu_*\{f > t\}$ :  $\{f > t\}$  and  $\{f > t\}$  are in  $A$ . Hence any bounded  $f \in L$  can be approximated uniformly by  $A$ -measurable step functions, and thus

$$L \subseteq L_1(A, \mu), \text{ with } \bar{\mu}(f) = \int f d\mu \text{ for } f \in L.$$

Conversely properties (1) to (4) imply also that

$$\forall f, \forall (f_n)_{n \in \mathbb{N}}, [\bar{\mu}(f_n) > -\infty, \bar{\mu}(\{f_n < f - \epsilon\}) \xrightarrow[n \rightarrow \infty]{} 0 \forall \epsilon > 0],$$

$$\limsup_{n \rightarrow \infty} \sup_{k \geq n} \bar{\mu}[(f_k - f_n)^+] = 0 \Rightarrow \lim_{n \rightarrow \infty} \bar{\mu}(f_n) > \bar{\mu}(f),$$

and hence that, if  $f_n$  is a Cauchy sequence in  $L$  converging in  $\bar{\mu}$ -measure to  $f$ , then  $f \in L$  is the norm limit of  $f_n$ . In particular, choosing  $f \in L_1(A, \mu)$  and  $f_n$  step functions, one obtains  $L = L_1(A, \mu)$ . One concludes now easily that  $\bar{\mu}$  is at least as good as the finitely additive integral:  $\forall f, \bar{\mu}(f) \leq \int^* f d\mu$ . Obviously,  $L = L_1(A, \mu)$  contains both the cylindrically integrable functions and the bounded continuous functions on  $\bar{E}$ .

Of course, one could still get conceivably more integrable functions by refining  $\bar{\mu}$  - for instance if one could prove  $\tau$ -smoothness of  $\mu$  on  $\bar{E}$ , one could use its regular extension to the Borel sets of  $\bar{E}$  for defining  $\bar{\mu}$ ; or one could try to get a lower  $\bar{f}$ , for instance by restricting the  $y \in V_\alpha(x)$  to be of essentially minimal norm.

#### 4.2 Using More Smooth Cylinder Measures

By Theorem 1, the invariant cylinder measures corresponding to different pairs  $(m, \sigma)$  are mutually singular. Thus the integral of a function - and even its integrability - may depend in a highly irregular way on the pair  $(m, \sigma)$ . To smooth this out, one could choose  $m$  and  $\sigma$  by some probability distribution  $P(m, \sigma)$ . Since the correspondence preserves convolution, and because of the idea that in some sense the sum of two independent random elements of  $B(I, C)$  is a fortiori random, one should certainly take  $P$  absolutely continuous with respect to

Lebesgue measure, and in some sense invariant under convolution. Since

$$\begin{aligned} \int D_X^\lambda(\tilde{x}) dQ(x) &= \int D_{m+\sigma X}^\lambda(x) dP(m, \sigma) dQ_0(x) \\ &= \int D_{(m\lambda+\sigma\lambda X)}^1(\tilde{x}) dP(m, \sigma) dQ_0(x) \\ &= \int D_{m+\sigma X}^1(x) dP^\lambda(m, \sigma) dQ_0(x) \end{aligned}$$

we see that for defining the value, we consider integrals of a fixed function with respect to the distribution  $P^\lambda$  of  $(\lambda m, \lambda \sigma)$  (where  $(m, \sigma)$  is  $P$ -distributed), and let the scale factor  $\lambda$  go to  $\infty$ . Asking that this family  $P^\lambda$  be invariant under convolution is asking that  $P$  be stable. This leads to choosing  $m$  and  $\sigma$  independently,  $m$  with the symmetric stable distribution of index  $\alpha$ , and  $\sigma$  with the one-sided stable distribution of index  $\beta$  (thus  $\alpha < 1$ ). The lim sup of the (upper-) integrals when the scale factor  $\lambda$  goes to  $\infty$  is then clearly decreasing when  $\alpha + 0$ , since  $\beta < \alpha$  the stable distribution with index  $\beta$  can be viewed as a mixture of stable distributions with index  $\alpha$  (choosing their scale factors according to the stable one-sided distribution with index  $\beta/\alpha$ ).

One is thus led to a formulation of the following type:

. For any bounded measurable function  $f$  on  $R_+$ , let

$$p(f) = \lim_{\alpha \rightarrow 0} \limsup_{\lambda \rightarrow \infty} \int_0^\infty f(\lambda x) dP^\alpha(x) ,$$

where  $P^\alpha$  is a one-sided stable distribution with index  $\alpha$  (its scale factor does not matter). Let  $\bar{\mu}$  denote a suitable extension (cfr. 4.1) Aumann and Shapley [1974]) of the invariant cylinder measure where  $m$

and  $\sigma$  are chosen independently with stable distributions of index  $\alpha$  ( $\alpha < 1$ ) - symmetric for  $m$  and one-sided for  $\sigma$ .

• If  $f$  is a function of several variables, let  $p_x(f)$  denote  $p$  of the function of a real variable  $x$  obtained by holding all variables but  $x$  fixed in  $f$ . Similarly  $\bar{p}_x(\Phi)$  will indicate that all variables but  $x$  are held fixed in  $\Phi$ .

• Let  $\Phi_v(x) = p_\lambda(\bar{p}_x(D_x^\lambda(x)))$ , and  $\Phi_{-v}(x) = -p_\lambda(\bar{p}_x(-D_x^\lambda(x)))$ : then  $v$  has a value  $\Phi_v$  if  $\Phi_v = \Phi_{-v}$  and is additive.

### 2.3 Reversing more limits and integrals?

As a general rule, one gets functionals with a larger domain by averaging before going to limits rather than after - in our context, this was already illustrated in "Values and Derivatives." Now the  $v$  appearing in the above formula for  $\Phi_v$  is not the given game, but obtained from it by the operator  $\psi$  of Section 1, which itself involves both averaging and limit operations.

Let us first show how  $\psi$  could be replaced by an operator - say  $\tilde{\psi}$  - where the averaging occurs before the limit operations, in order to remove as much as possible the basic restriction that we can talk only of games that have an extension in some sense.

Remark first, that it is sufficient to compute  $w(x) = \psi(v)(x)$  for step functions  $x$  - either because  $v(w)[x - x'] < \|x' - x\| \cdot \|w\|$  implies (for  $\|w\| < \infty$ ) that  $w$  can anyway be uniquely extended to  $B(I,C)$ , or using the fact that the cylinder measures on  $B(I,C)$  are also cylinder measures on the space of step functions  $\in B(I,C)$ .

We return to the basic idea underlying the proof in Mertens [1981] of both theorem B and its application to the extension of games - that was used in Section 1.

Let  $x$  denote a step function, and let  $\pi$  denote a finite measurable partition such that  $x$  is constant on every element of  $\pi$ .

Given any vector  $v$  of non atomic elements of FA, let, for any  $A \in \pi$ ,  $O_t^A$  denote an increasing family of measurable subsets of  $A$  with  $v(O_t^A) = tv(A)$  ( $\forall t : 0 < t < 1$ ) (and with  $O_0^A = \emptyset$ ,  $O_1^A = A$ ).

For any  $n > 0$ , for any permutation  $\sigma$  of  $\{1, \dots, n\}$ , and for any  $\epsilon \in \{-1, 1\}^n$ , define  $x_0^i = O_{i/n}^A \setminus O_{(i-1)/n}^A$ , and, denoting  $(\sigma, \epsilon)$  by  $\omega$ , let  $O_t^{A, \omega}$  be defined by  $(\bigcup_{i \leq nt} x_0^{\sigma(i)}) \cup B_{\sigma([nt]+1)}$  where  $[x]$  denotes the integer part of  $x$ , and  $B_k = O_{((k-1)/n)+t_n}^A \setminus O_{(k-1)/n}^A$  if  $\epsilon_k = 1$ ,  $B_k = O_{k/n}^A \setminus O_{(k/n)-t_n}^A$  if  $\epsilon_k = -1$  - and where  $t_n = t - [nt]/n$ .

Then, for every  $\omega$  and  $t$  we still have  $v(O_t^{A, \omega}) = tv(A)$ , and if  $\omega$  is chosen at random, we have for all  $x \in A$ ,  $P(x \in O_t^{A, \omega}) - t \leq 1/2n$ .

(The  $\epsilon$  is not strictly necessary, it is just introduced to preserve the symmetry with the opposite order.)

Let now, for any  $\pi$ , and any collection  $O^\pi, v$  of such increasing families  $(O_t^A)_{\substack{A \in \pi \\ 0 \leq t \leq 1}}$ , and any  $n$ ,  $\Omega_n$  denote the finite probability space where independently for each  $A \in \pi$ , some  $\omega = \omega_A$  is chosen at random.

Also, for any  $\pi$ -measurable ideal set function  $x$ , and any  $\omega \in \Omega_n$ , let  $x_\omega = \bigcup_{A \in \pi} O_{x(A)}^{A, \omega}$ : then  $x^1 < x^2 \Rightarrow x_\omega^1 < x_\omega^2$ ,  $v(x_\omega) = v(x) \forall x$ , and  $E(x_\omega) - x \leq 1/2n$ .

Let also, for a general  $\pi$ -measurable function  $x$ ,

$$x_\omega = [\text{Max}(0, \text{Min}(1, x))]_\omega .$$

Define now

$$\psi_{\pi, 0^\pi, v, n}^{\tau, v}(x) = \frac{1}{2\tau} \int_0^1 (E_{\Omega(n!)}(v[(t + \tau x)_\omega] - v[(t - \tau x)_\omega])) dt$$

(where  $v$  still denotes the constant sum game corresponding to the originally given game).

For any given  $\pi$ , and any vector  $v$ , denote by  $F_{v, \pi}$  the set of all possible families  $0^\pi, v$ . For any given  $\pi$ , the  $F_{v, \pi}$  form a filter  $F_\pi$ , when  $v$  ranges over the increasing filtering set of finite subsets of the nonatomic elements of FA.

Similarly the partitions  $\pi$  can be ordered by refinement. Then  $\lim_{\tau \rightarrow 0} \lim_{\pi} \lim_{n \rightarrow \infty} \psi_{\pi, 0^\pi, v, n}^{\tau, v}(x) = \psi^v(x)$  should be the analog of our  $\psi$  from Section 1 but with all limits done after any averaging.

More formally, define a filter  $F$  on 4-tuples  $(\tau, \pi, 0^\pi, v, n)$  (more formally on  $(R \times (\bigcup_\pi F_{\phi, \pi}) \times N)$ ) by  $F \in F$  if  $\exists \epsilon : \forall \tau : 0 < |\tau| < \epsilon \exists \pi_o : \forall \pi > \pi_o, \exists v (= (v_1 \dots v_k)) : \forall 0^\pi, v \in F_{v, \pi} \exists n_o : \forall n > n_o (\tau, \pi, 0^\pi, v, n) \in F$ .

Then, we define  $\psi$  by  $v \in \text{Dom}(\psi) \equiv [\lim_{F} \psi_{\pi, 0^\pi, v, n}^{\tau, v}(x)]$  exists for any step function  $x \Rightarrow [\psi(v)](x) = \lim_{F} \psi_{\pi, 0^\pi, v, n}^{\tau, v}(x) \forall x$  step function.

Obviously  $\text{Dom}(\psi)$  is a closed (using  $\|\psi\| = 1$ ) symmetric space, and  $\psi$  is a positive linear symmetric operator on  $\text{Dom}(\psi)$ . Further  $\|\psi\| = 1$  - this follows from completely similar computations as those in Section 1, and is the main point where the specific structure of the  $0^\pi, v(x < x' \Rightarrow x_\omega < x'_\omega \forall \omega)$  is used. Similarly one gets, under mild

continuity assumptions on  $v$  at  $\emptyset$  and  $I$ , that  $[\psi(v)](I) = v(I)$ .

One could thus use this  $\psi$  followed by the operation described in the previous section. However it is now tempting - and possible - to put all averagings before any limit operation.

But to do this, one may want to consider an alternative to integrating with respect to an appropriate extension of the (finitely additive) cylinder measure - in order to sidestep the difficulties of finitely additive integration theory (in what concerns the integrability of functions, and in what concerns changing the order of integration and the permutation of limits and integrals - although my old paper<sup>4/</sup> helps a good way for those last two questions).

The cylinder measure  $Q$  can be obtained - as shown in the proof of Theorem 1 - in the following way: first select  $m$  and  $\sigma$  at random according to  $P$ , next, for any partition  $\pi$ , select independently on each partition element the (constant) value of  $x$  on that partition element as a Cauchy  $(m, \sigma)$  random variable. This gives an approximation  $Q_\pi$  to  $Q$ , that converges weakly to  $Q$  on  $E$  when  $\pi$  is refined.  $Q_\pi$  is a (countably additive) probability carried by the finite dimensional subspace of  $B(I, C)$  of all  $\pi$ -measurable step functions.

Now the operator  $D$  and the averaging for  $Q_\pi$  can without problem be pushed before the  $\lim_F$ , together with all other averagings and then is no integrability problem at least if  $v$  is of bounded variation. On the other hand the limit over all refinements of  $\pi$  is best retained after the  $\lim_F$  has been done (and before the  $\lim_\lambda$  over  $\lambda(p_\lambda)$ ).

This was just to point out that the formulation adopted in this paper is by no means unique or optimal - and that in particular one could to some extent dispense altogether with the assumption that the game has an extension. It was adopted chiefly for expository reasons.

Certainly a lot remains to be done - i.e., convincing theorems - to get a good formulation.

Footnotes

- 1/ The equality  $\gamma_0(x) = P(\bigcup_i J_i \text{ has density 0 at } x)$  has to be proved only when  $a_0 > 0$ : we claim that a.s. on  $x \notin \bigcup_i J_i$ , this set has density zero at  $x$ . It is indeed sufficient to prove this conditionally to the set of atoms and the fraction of  $a_0$  coming after  $x$  (or before) - which reduces (by renormalization) the problem to the case  $x = 0$ . Let  $X_t = (1/t) \sum_{i>1} (a_i/a_0) \wedge t I(z_i < t)$ :  $X_t$  is an upper bound for the density of  $\bigcup_i J_i$  up to time  $t$ , so it is sufficient to show that  $X_t \rightarrow 0$  a.s. But, if  $F_t$  denotes the  $\sigma$ -field generated by all variables  $t : Z_i$ , then, when reversing the usual order on the time interval  $[0,1]$ ,  $X_t$  becomes a positive supermartingale w.r.t.  $F_t$ , whose expectation goes to zero as we have seen: thus  $X_t$  goes to zero a.s. We would like to stress that this equality cannot be dispensed within a random order approach: indeed, if  $f(x) = I(x > q)$ , and if some player of the ocean pivots, since he is negligible, it is in fact the infinitesimal coalition  $ds$  that immediately follows him that pivots - so to impute this event to the credit of the ocean, one needs that this infinitesimal coalition consists essentially only of oceanic players - i.e., the ocean must have density 1 to the right of  $q$ . The same applies to the left of  $q$  if  $f(x) = I(x > q)$ .
- 2/ That is, for any  $P$  in the right hand member, there exists a unique cylinder measure with this Fourier transform, and this cylinder measure is invariant.
- 3/ "Measurability in a Banach space" Indiana University Mathematics Journal, Vol. 26, No. 4, pp. 663-677 [1977].
- 4/ J. F. Mertens: "Integration des mesures non denombrablement additives: une generalisation du lemme de Fatou et du theoreme de convergence de Lebesgue", Annales de la Societe Scientifique de Bruxelles, t. 84, 88, 231-239 [1970].

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